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## Topological Privacy

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# Topological Privacy: Lattice Structures and Information Bubbles for Inference and Obfuscation

Final Report to AFOSR  
Award FA9550-14-1-0012

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## Abstract

Information has intrinsic geometric and topological structure, arising from relative relationships beyond absolute values or types. For instance, the fact that two people did or did not share a meal describes a relationship independent of the meal's ingredients. Such relationships give rise to lattices. Lattices have topology. That topology informs the ways in which information may be observed, hidden, inferred, and dissembled. Privacy preservation may be understood as finding isotropic topologies, in which relationships appear homogeneous. Moreover, the underlying lattice structure of those topologies has a temporal aspect, which reveals how isotropy may degrade over time, thereby puncturing privacy.

Dowker's Theorem establishes a homotopy equivalence between two simplicial complexes derived from a relation. From a privacy perspective, one complex describes individuals with common attributes, the other describes attributes shared by individuals. The homotopy equivalence is an alignment of certain common cores of those complexes, effectively interpreting sets of individuals as sets of attributes, and vice-versa. That common core has a lattice structure. An element in the lattice consists of two components, one being a set of individuals, the other being an equivalent set of attributes. The lattice operations join and meet each amount to set intersection in one component and set union followed by a potentially privacy-puncturing inference in the other component.

One objective of this research has been to understand the topology of the Dowker complexes, from a privacy perspective. First, privacy loss appears as simplicial collapse of free faces. The actual collapse is local, but the property of fully preserving both attribute and association privacy requires a global condition: a particular kind of spherical hole. Second, by looking at the link of an individual in its encompassing Dowker complex, one can characterize that individual's privacy via another sphere condition. That characterization generalizes to group privacy. Third, even when long-term privacy is impossible, homology provides lower bounds on how an individual may defer identification, when that individual has control over how to reveal attributes. Intuitively, the idea is to first reveal information that could otherwise be inferred. This last result in particular highlights privacy as a dynamic process. Privacy loss may be cast as gradient flow. Harmonic flow for privacy preservation may be fertile ground for future research.

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# 1 Introduction

Privacy is the ability to control how much an individual or entity reveals about itself to others. Fundamental research into privacy seeks to understand the limits of that ability.

A brief history of privacy should include the following:

- **The right** to privacy as a legal principle, appearing in an 1890 Harvard Law Review article [20]. The article was a reaction to the then modern technology of photography and the dissemination of gossip via print media.
- **A demonstration** linking supposedly anonymous information with public data, thereby revealing sensitive information [17]. The demonstration employed birth date, gender, and zip code to link anonymous public insurance information with voter registration data. Doing so produced the health record of the governor of Massachusetts. This privacy failure suggested a first form of homogenization, called *k-anonymity*. Roughly, the idea was to structure databases in such a way that a database could respond to any query with an answer consisting of no fewer than  $k$  individuals matching the query parameters.
- **The discovery** that it is impossible to preserve the privacy of an individual for even a single attribute in the face of repeated statistical queries over a population [2], *unless* answers to those queries are purposefully perturbed with noise of magnitude on the order of at least  $\sqrt{n}$ . Here  $n$  is the size of the population. The significance of this discovery is to underscore how difficult it is to preserve privacy while retaining information utility.
- **Netflix Prize**. In 2006, Netflix offered a \$1M prize for an algorithm that would predict viewer preferences better than Netflix’s internal algorithm. Netflix made available some of its historical user preferences, in anonymized form, as a basis for the competition. Once again, it turned out that one could link this anonymized data with other publicly available databases, resulting in the potential (and in some cases actual) identification of Netflix viewers and their entire viewing history [15]. Whereas in the earlier health example, a few specific observables made linking possible (global coordinates, one might say, namely birth date, gender, zip code), in the Netflix example, the intrinsic geometric structure of the database facilitated linking via a wide variety of observables (local landmarks, one might say, namely movies that were characteristic for each individual). Key was sparsity of information: 8 movie ratings and dates were generally enough to uniquely characterize 99% of viewers in the Netflix Prize dataset, even with errors in the ratings and dates.
- **Differential Privacy** [5, 4] seeks to avoid the previous privacy failures by focusing on local rather than absolute privacy guarantees. The underlying approach in differential privacy is for a database to answer statistical queries with a particular stochastic blurring. Specifically, the probability that an interrogator of the database will make any particular inference should depend only in a very small way on whether any one individual does or does not have a particular attribute (such as even being in the database). We might call this *stochastic homogeneity*.
- **Randomized Response**. Differential privacy is further significant because it makes explicit the dynamic nature of privacy; there may be no enduring privacy guarantees but



there are differential guarantees. A particular form is *randomized response*, a technique used in the social sciences to elicit reliable aggregate answers to sensitive questions, asking the question of many people, but perturbing individual answers stochastically so as not to learn much about any one individual [19]. A version has been employed by Google to find malware [8]. (We note a form of ergodicity: the averaging that would destroy privacy for an individual with repeated queries over time allows for utility of information at any instant in time over a large population.)

Privacy has both a combinatorial component and a statistical component. Prior research has largely focused on statistical techniques, both to preserve privacy and to puncture privacy. One of the goals of this research is to understand the combinatorial component of privacy, leading naturally to methods from combinatorial topology.

A desire to understand the geometry and topology of the types of inferences revealed by the Netflix Prize formed the specific motivation for our research initially. Subsequently, we realized that the lattice structure found in that geometry had broader applicability, providing an ability to model the dynamics of privacy more generally.

## 2 Outline

The remaining sections and appendices present the following material:

- 3: Toy examples illustrating how a relation may lead to privacy loss in the presence of background information. This section also introduces the *doubly-labeled poset* associated with a relation, to model such inferences. The elements of the poset are pairs, each a set of individuals and a set of attributes.
- 4: Formal description of the *Galois Connection* associated with a relation. The section first defines, for any relation, two simplicial complexes called *Dowker complexes*. One complex represents sets of individuals with shared attributes, the other represents sets of attributes shared by individuals. The Galois Connection then establishes a homotopy equivalence between the Dowker complexes, thereby generating the relation's doubly-labeled poset. The homotopy equivalence gives rise to closure operators, with "closure" in the poset modeling inference of unobserved attributes from observed attributes (or unobserved individuals from observed individuals). The section defines *attribute privacy* and *association privacy*.
- 5: A characterization of privacy in terms of the absence of free faces in the relevant Dowker complex. This section observes as well that the only connected relations able to preserve both attribute and association privacy must look like either linear cycles or boundary complexes. In particular, the number of individuals and attributes must be the same.
- 6: Conditional relations as models for simplicial links. A conditional relation is much like a conditional probability distribution. It might, for instance, represent the possible arrangement of remaining attributes among individuals, after some attributes have already been observed.
- 7: A characterization of individual and group privacy in terms of spherical and boundary complexes for the relation that models the individual's or group's link in its Dowker complex.
- 8: A brief exploration of holes in relations, focusing on attribute spaces generated by bits.
- 9: A small example exploring the possibility of increasing privacy by change-of-coordinate transformations.
- 10: A lengthy exploration of how someone can delay identification by releasing attributes selectively in a particular order. This idea leads to the notion of *informative attribute release sequences*, how to find such sequences in the Galois lattice, and the value of homology as a lower bound for the number and length of such sequences.
- 11: Computation of the homology and maximal informative attribute release sequences present in two relations found on the world wide web. One relation describes Olympic athletes and their medals, the other describes jazz musicians and their bands.

- 12:** A more general perspective of inference as motion in lattices, not necessarily directly derived from a relation. This perspective suggests connections to randomized response techniques.
- 13:** An examination of the ability to obfuscate strategies and/or goals in graphs where motions may be nondeterministic or stochastic.
- 14:** A possible category for representing relations, along with an analysis of morphism properties. The morphisms between relations in this category induce simplicial and therefore continuous maps on the Dowker complexes. This section shows how a surjective morphism at the set level generates the image lattice via lattice operations performed on images of certain elements from the domain lattice.
- A:** A summary of the basic notation and definitions used in this report.
- B:** A summary of the basic tools used in this report, establishing the homotopy equivalences and closure operators mentioned above.
- C:** Construction of links and deletions, and examination of the privacy properties each inherits from its encompassing relation. This section explores the significance of free faces in the Dowker complexes. The section further proves that a relation with more attributes than individuals cannot preserve attribute privacy.
- D:** Proof that the problem of finding a minimal set of attributes from which another attribute may be inferred is *NP*-complete. This stands in contrast to the observation that the problem of finding *some* set of attributes from which another may be inferred (or reporting that no such set exists) is computable in polynomial time.
- E:** Detailed proofs of the results claimed in Section 7. Also a detailed proof of the assertion from Section 5 regarding relations that preserve both attribute and association privacy.
- F:** Detailed proofs of the connection between maximal chains in the Galois lattice and informative attribute release sequences. When such sequences are order-independent they correspond to spherical holes, leading to the concept of an *isotropic* sequence.
- G:** Detailed proof that homology establishes a lower bound for the number and length of maximal chains in a relation's Galois lattice, and thus for the number and length of informative attribute release sequences that may be used to delay identification.
- H:** An application of the previous results with the aim of obfuscating the identification of strategies for attaining goals in graphs with uncertain transitions.
- I:** Detailed proofs of the assertions of Section 14 regarding morphisms.
- J:** Some additional examples:
  1. Dunce Hat: modeled as a relation for which the Dowker attribute complex is contractible but has no free attribute faces, meaning the relation preserves attribute privacy.

2. Disinformation: An example that glues together two copies of the Möbius strip, thereby removing free faces and creating a form of homogeneity that preserves attribute privacy yet retains the utility of identifiability.
3. Insufficient Representation: If there are insufficiently many individuals in a relation generated by bits, attribute inference is possible.
4. A Matching Example: When many individuals are being observed, cardinality constraints allow for inferences beyond those discussed in this report. One can model some such inferences using links and joins. We have not reported that work here, merely provide one example.

## List of Primary Symbols

<u>Symbol</u>	<u>Typical Meaning</u>	<u>Page(s)</u>
$X$	space of individuals	16, 85
$Y$	space of attributes	16, 85
$R$	relation on $X \times Y$	16, 85
$X_y$	individuals with attribute $y$ (usually in the context of relation $R$ )	16, 85
$Y_x$	attributes of individual $x$ (usually in the context of relation $R$ )	16, 85
$Q$	another relation, often representing a link in a complex	27, 90, 43
$\Sigma, \Gamma$	generic simplicial complexes	81
$\Psi_R$	complex generated by sets of individuals with a common attribute	16, 85
$\Phi_R$	complex generated by sets of attributes shared by some individual	16, 85
$\sigma$	usually a simplex representing individuals in $\Psi_R$	
$\gamma$	usually a simplex representing attributes in $\Phi_R$	
$\phi_R$	homotopy equivalence from sets of individuals to shared attributes	17, 85
$\psi_R$	homotopy equivalence from sets of attributes to sharing individuals	17, 85
$P$	partially ordered set (poset)	83
$\mathfrak{F}(\Sigma)$	face poset of the simplicial complex $\Sigma$	17, 83
$\Delta(P)$	order complex of the poset $P$	18, 83
$P_R$	doubly-labeled poset associated with relation $R$	14, 20, 86
$L$	(inference) lattice	(61) 84
$P_R^+$	Galois lattice formed from $P_R$	38
$\{(\sigma_k, \gamma_k) < \cdots < (\sigma_0, \gamma_0)\}$	chain of length $k$ in the lattice $P_R^+$	44, 109, 83
$y_1, \dots, y_k$	informative attribute release sequence (for a relation $R$ )	41
$V$	set of vertices in a simplicial complex or in a graph	
$\partial(V)$	simplicial boundary complex with vertices $V$	24, 82
$\mathbb{S}^{-1}$	sphere of dimension $-1$ , modeling the empty complex $\{\emptyset\}$	81
$\mathbb{S}^1$	circle	24
$\mathbb{S}^{n-2}$	sphere of dimension $n-2$	24, 82
$C_k(\Sigma; \mathbb{Z})$	group of simplicial $k$ -chains over $\Sigma$ , with integer coefficients	81
$\tilde{\partial}$	(family of) reduced boundary map(s) $C_k(\Sigma; \mathbb{Z}) \rightarrow C_{k-1}(\Sigma; \mathbb{Z})$	82
$\tilde{H}_k(\Sigma; \mathbb{Z})$	reduced $k$ -dimensional homology group of $\Sigma$ , with integer coefficients	82
$G$	nondeterministic or stochastic graph	65, 67
$\Delta_G$	strategy complex of a graph	66, 67
$\overline{\Delta}_G$	source complex of a graph	119
$\simeq$	homotopy equivalence	83
$*$	simplicial join	83
$\vee$	either topological wedge sum or lattice join	83, 84
$\wedge$	lattice meet	84

### 3 Privacy: Relations and Partially Ordered Sets

Our investigation of privacy in this report will be in terms of relations. As we will see in this section and the next, relations give rise to simplicial complexes, which give rise to partially ordered sets, which expose an underlying lattice structure. That lattice structure makes explicit how privacy may be preserved or lost through so-called *background knowledge*. As we will see in Section 10, the lattice structure also makes explicit how identification may be delayed by careful release of information.

#### 3.1 A Toy Example: Health Data and Attribute Privacy

Consider the following relation  $H$ , describing the results of a health study for four patients and three attributes. The patients have been anonymized and are represented simply by the set of numbers  $\{1, 2, 3, 4\}$ . The three attributes are drawn from the set  $\{\text{SMOKES}, \text{HAS\_CANCER}, \text{DRINKS\_SODA}\}$ .

One can describe a relation equivalently either as a matrix or as a set of pairs:

Relation  $H$  as a matrix:

$H$	SMOKES	HAS_CANCER	DRINKS_SODA
1	•	•	
2		•	•
3			•
4			•

Relation  $H$  as a set of pairs:

$$\{(1, \text{SMOKES}), (1, \text{HAS\_CANCER}), (2, \text{HAS\_CANCER}), (2, \text{DRINKS\_SODA}), \\ (3, \text{DRINKS\_SODA}), (4, \text{DRINKS\_SODA})\}.$$

#### Assumptions

Before discussing privacy further, we make some assumptions that hold throughout the report:

**Assumption of Relational Completeness:** We generally assume that a *relation is complete*, meaning it is not missing any elements (a relation could contain extra elements, which may be useful as disinformation). For example, if we observe that someone drinks soda and has cancer in relation  $H$ , then we would conclude that we are observing individual #2. We would be surprised to see that individual smoke. If for some reason we ever do see the individual smoke, then we would deem our observations to be *inconsistent* with relation  $H$ . The meaning of inconsistency depends on context. At top-level it may mean that the relation or observation is errorful. When making conditional observations, an inconsistency may actually supply useful information, as we will see in Lemma 11 on page 29.

**Assumption of Observational Monotonicity:** Even though we assume relations are complete, we do *not* assume that observations are complete. Instead we assume: *Observing that an individual has a particular attribute is meaningful; lack of such an observation does not necessarily imply that an individual fails to have the unobserved attribute*. The motivation

for this assumption is that one may yet discover that the individual has the attribute. For example, suppose we observe someone (whom we know to be part of relation  $H$ ) drinking soda. Even if that is all we observe, we do *not* conclude that the individual is cancer free. It could be that we might yet observe the individual to have cancer.

If absence of an attribute is significant *and* that absence is measurable, then both the attribute and its negation could and perhaps should appear explicitly in the relation as distinct mutually exclusive attributes. For instance, PRIME versus COMPOSITE might be such a pair of attributes for integers greater than 1.

**Assumption of Observational Accuracy:** We assume that *observations are accurate*. For instance, if we observe an integer to be either PRIME or COMPOSITE, then we do so correctly.

**Comments:** The three assumptions above are *desiderata* for how the mathematical abstractions of this report fit into the real world. Some comments are in order:

- In and of itself, a relation defines a particular kind of world, a bipartite graph, and there is no need for something like a completeness assumption.
- The monotonicity and accuracy assumptions then describe a sensor for that world and how to interpret observations.

The purpose of the assumptions in the real world is largely to ensure consistency between different relations and with observations.

- The monotonicity assumption is important because information generally aggregates asynchronously. Together with the other assumptions, this assumption means that one may view relations as monotone Boolean functions, and thus may leverage methods from combinatorial topology.
- One may incorporate errors into the relational and observational models, for instance by blurring a relation. For very large integers, a relation might allow some integers to have *both* PRIME and COMPOSITE as attributes. Although an integer is one or the other, the relation admits to uncertainty by allowing both attributes at once. Indeed, some relations purposefully introduce such blurring to preserve privacy. And, in robotics, relational blurring in a sensor-compatible fashion can be a useful technique for establishing the topology of a region, for instance when dualizing sensors and landmarks [10].

## Privacy Implications

If the health study  $H$  is publicly available, then it has the following privacy implications:

- Suppose someone named Bob tells his friend Alice that he was part of the study. Alice knows that Bob smokes everywhere he goes, so she can infer that he is Patient #1 and has cancer. (This is an example of inference in a relation using background knowledge.)
- Suppose Cindy is Patient #2. She has full privacy as far as relation  $H$  is concerned. In particular, as we saw already, Cindy can tell her friends that she was part of the health study while drinking soda and those friends will not be able to conclude that she has cancer.

- Patients #3 and #4 are not only indistinguishable from each other but also from Cindy (patient #2). This is a very strong form of anonymity. Even if one of them reveals that s/he drinks soda, s/he will remain indistinguishable from the other two patients who drink soda.

**Caveat:** In the last case, if Cindy reveals that she has cancer and is seen to be different from the other individuals, then one may be able to remove her from the relation, narrowing the focus and creating a new relation that may allow additional inferences. Similar caveats hold for the other bullets. Deletions are discussed further in Appendix C.

**Modifying a Relation to Increase Privacy** We can make a small change in relation  $H$  that enhances privacy. If we artificially give patient #3 the attribute SMOKES, then we obtain the following modified relation  $H'$ :

$H'$	SMOKES	HAS_CANCER	DRINKS_SODA
1	•	•	
2		•	•
3	•		•
4			•

Now Bob may reveal to Alice that he was part of the health study without Alice being able to infer that he has cancer, even though she knows that everyone knows that he smokes. In fact, more generally, one can no longer infer cancer from smoking.

Such an artificial entry<sup>1</sup> in the relation is a form of *disinformation*. It certainly skews statistics and utility. It also increases privacy.

---

<sup>1</sup>Terminology: We often use the term '**entry**' to mean an element of a relation, as in a matrix.



### 3.2 A Dual Perspective: Payroll Data and Association Privacy

The previous example examined a relation from the perspective of *attribute privacy*: we were interested in understanding how observation of some attribute(s) implies other attribute(s). A dual perspective is *association privacy*, in which one seeks to understand how some associations between individuals imply others.

The following “salary” relation  $S$  has the same matrix structure as  $H$  did earlier, but with different semantics. This relation represents employees {Bob, Mary, Frank, Julie} working on secret projects {a, b, c}. Now the employee names are visible so that a payroll clerk can disburse salaries correctly, but the actual projects are anonymous.

$S$	a	b	c
Bob	•	•	
Mary		•	•
Frank			•
Julie			•

The salary relation  $S$  facilitates the following implications regarding individuals:

- If someone tells the payroll clerk that Julie is the lead of a very important project, then the payroll clerk can infer that Mary and Frank may have valuable information.
- In contrast, if someone tells the payroll clerk that Bob is the lead of a very important project, the payroll clerk cannot be sure that Mary is also working on that project.

Regarding disinformation: Observe how adding the artificial entry (Julie, a) prevents the payroll clerk from inferring that Mary and Frank have valuable information, even if the payroll clerk knows that Julie does:

$S'$	a	b	c
Bob	•	•	
Mary		•	•
Frank			•
Julie	•		•

### 3.3 Privacy Preservation and Loss: A Poset Model

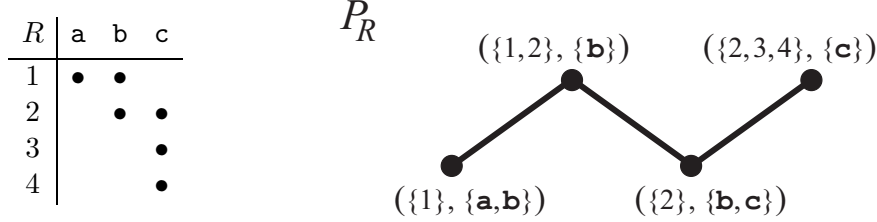


Figure 1: Relation  $R$  serves as a model for the two examples of Sections 3.1 and 3.2. The doubly-labeled poset  $P_R$  describes the inferences facilitated by  $R$ .

Figure 1 shows a relation  $R$  that serves as a model for both the health example of Section 3.1 and the payroll example of Section 3.2. The relation is identical to those given earlier, but with abstract labels in place of both individuals and attributes. The figure also depicts a partially ordered set (poset)  $P_R$ , designed to model the inferences discussed previously. We refer to that poset as the *doubly-labeled poset associated with  $R$* . We next discuss the semantics of  $P_R$ . Section 4 discusses the construction of  $P_R$ . The underlying concepts are important throughout the report.

#### Semantics of the poset $P_R$ :

- Each element in the poset consists of a pair  $(\sigma, \gamma)$ , with  $\emptyset \neq \sigma \subseteq \{1, 2, 3, 4\}$  describing a set of individuals and  $\emptyset \neq \gamma \subseteq \{a, b, c\}$  describing a set of attributes. We say that the poset element is *labeled with  $\sigma$  and  $\gamma$* . The meaning of such a double-labeling is:
  - (a) All individuals in  $\sigma$  have all attributes in  $\gamma$ .
  - (b) If an individual has at least all the attributes in  $\gamma$ , then that individual must be in  $\sigma$ . For example, we see that individual #2 and only individual #2 has both attributes  $b$  and  $c$  in  $R$ .
  - (c) If an attribute is shared by at least all individuals in  $\sigma$ , then that attribute must be in  $\gamma$ . For example, attribute  $b$  and only attribute  $b$  is shared by both individuals #1 and #2.
- The partial order for  $P_R$  is described by the edges in the figure. There is an edge between two elements  $(\sigma_1, \gamma_1)$  and  $(\sigma_2, \gamma_2)$  of  $P_R$  whenever the corresponding sets are subset comparable. In particular,  $(\sigma_1, \gamma_1) \leq (\sigma_2, \gamma_2)$  in  $P_R$  precisely when  $\sigma_1 \subseteq \sigma_2$  and  $\gamma_1 \supseteq \gamma_2$ . [Observe that the comparability ( $\subseteq$  versus  $\supseteq$ ) is opposite for  $\sigma$  versus  $\gamma$ .]

#### Using the poset $P_R$ for attribute inference:

Suppose  $\gamma$  is *any* nonempty subset of attributes in  $\{a, b, c\}$ . Then:

- (i) Either: no individual has all the attributes  $\gamma$ . For example, no individual has both attributes  $\{a, c\}$ . We would not expect to see  $\gamma$  and so  $\gamma$  does not appear in the poset.

- (ii) Or:  $\gamma$  is a subset of at least one set of attributes that does appear in the poset. In this case, one *may* be able to enlarge  $\gamma$  nontrivially, resulting in privacy loss.

For example, imagine that individual #1 (Bob in our first example above) tells us that he has attribute **a** (SMOKES). So  $\gamma = \{\mathbf{a}\}$ . The poset then allows us to infer that Bob must also have attribute **b** (HAS\_CANCER). Why? Because  $\{\mathbf{a}, \mathbf{b}\}$  is a minimal set in  $P_R$  containing  $\{\mathbf{a}\}$ .

We can say yet more: The element labeled with  $\{\mathbf{a}, \mathbf{b}\}$  is also labeled with  $\{1\}$ . So now we know that Bob is individual #1.

Regardless of whether Bob ever actually talks to us, the poset tells us that individual #1 *could* suffer privacy loss, and in fact, is uniquely identifiable without needing to reveal everything about himself.

Similar reasoning is possible for **association inference**, as we saw earlier.

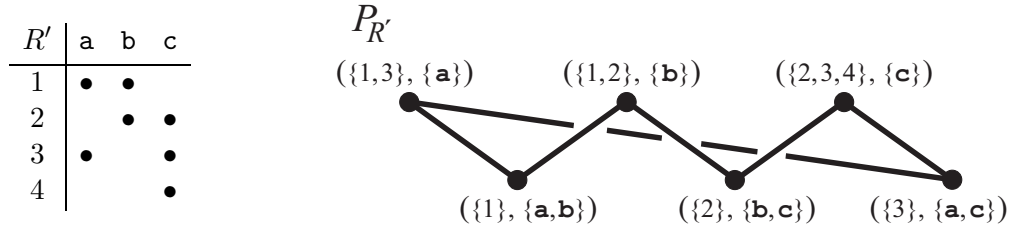


Figure 2: A relation  $R'$  along with its doubly-labeled poset  $P_{R'}$ . The relation preserves attribute privacy but allows a small amount of association inference: If one sees individual #4 in some context **c**, then one can infer that individuals #2 and #3 are also present in that same context, without needing to observe them directly.

**Disinformation Revisited:** Figure 2 shows relation  $R'$ , constructed from  $R$  by adding an entry of disinformation, much as we constructed  $H'$  from  $H$  earlier. The figure also shows the doubly-labeled poset  $P_{R'}$ . Observe that it is no longer possible to infer  $\{\mathbf{a}, \mathbf{b}\}$  from  $\{\mathbf{a}\}$  because  $\{\mathbf{a}\}$  now appears directly in the poset. The added entry has increased attribute privacy.

There is, however, still some opportunity for making association inferences. For instance, knowing that individual #4 (Julie earlier) works on an important secret project still allows the inference that individuals #2 and #3 might have valuable information. That is because the minimal set containing  $\{4\}$  in the poset is  $\{2, 3, 4\}$ . Notice that no such association inference is possible if someone says that individual #3 works on an important secret project, though that would have been possible in the original relation  $R$ .

## 4 The Galois Connection for Modeling Privacy

Section 3 showed by example how a relation determines a partially ordered set (poset) useful for modeling privacy. The elements in the poset are pairs — a set of attributes and a set of individuals — that are equivalent from the relation’s perspective. Privacy loss occurs when an observer has data (for example, background knowledge) that is not directly in the poset but is a proper subset of some set of attributes or individuals in the poset. The observer may then infer some additional attributes or individuals. This section develops the connection between relations and posets more precisely, continuing to use the earlier examples for illustration. See also Appendix B for additional material and notation.

### 4.1 Dowker Complexes

**Definition 1** (Dowker Complexes). *Let  $X$  and  $Y$  be finite discrete spaces and let  $R$  be a relation on  $X \times Y$ . This means  $R$  is a set of pairs  $(x, y)$ , with  $x \in X$  and  $y \in Y$ . We frequently view  $R$  as a matrix of 0s and 1s, or blank and nonblank entries, with  $X$  indexing rows and  $Y$  indexing columns.*

- (a) *We often refer to elements of  $X$  as individuals and to elements of  $Y$  as attributes.*
- (b) *For each  $x \in X$ , let  $Y_x = \{y \in Y \mid (x, y) \in R\}$ . Then  $Y_x$  consists of all attributes of individual  $x$ . We may view  $Y_x$  as a row of  $R$ . The row is blank if  $Y_x = \emptyset$ .*
- (c) *For each  $y \in Y$ , let  $X_y = \{x \in X \mid (x, y) \in R\}$ . Then  $X_y$  consists of all individuals who have attribute  $y$ . We may view  $X_y$  as a column of  $R$ . The column is blank if  $X_y = \emptyset$ .*
- (d) *We next define two simplicial complexes  $\Phi_R$  and  $\Psi_R$ :*

$$\begin{aligned}\Phi_R &= \{\gamma \subseteq Y \mid \text{there exists } x \in X \text{ such that } (x, y) \in R \text{ for all } y \in \gamma\}, \\ \Psi_R &= \{\sigma \subseteq X \mid \text{there exists } y \in Y \text{ such that } (x, y) \in R \text{ for all } x \in \sigma\}.\end{aligned}$$

*Special cases: If  $X$  and  $Y$  are both nonempty, then the empty simplex  $\emptyset$  is in both  $\Phi_R$  and  $\Psi_R$ . Otherwise, with some exceptions discussed later (Section 6, Section 10, and Appendix C), we take both complexes to be void.*

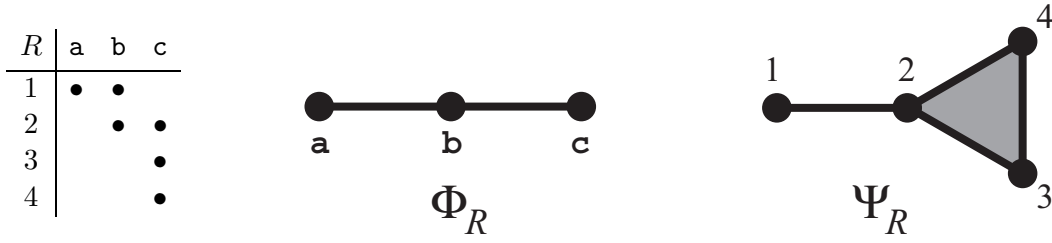
*We refer to  $\Phi_R$  and  $\Psi_R$  as Dowker complexes after the author of upcoming Theorem 2.*

*Interpretation: A nonempty set  $\gamma$  of attributes is a simplex in  $\Phi_R$  precisely when at least one individual has at least all the attributes in  $\gamma$ . We refer to any such individual as a witness for  $\gamma$ .*

*Similarly, a nonempty set  $\sigma$  of individuals is a simplex in  $\Psi_R$  precisely when there is at least one attribute that is shared by at least all the individuals in  $\sigma$ . We refer to any such attribute as a witness for  $\sigma$ .*

Figure 3 shows the Dowker complexes for the relation  $R$  of Section 3.3.

Dowker’s Theorem [3] says that the two simplicial complexes  $\Phi_R$  and  $\Psi_R$  have the same homotopy type. As we will see, the maps establishing that homotopy equivalence define the poset  $P_R$  and describe how privacy may be lost.

Figure 3: Simplicial complexes  $\Phi_R$  and  $\Psi_R$  associated with relation  $R$ .

**Theorem 2** (Dowker [3]). *Suppose  $R$  is a relation on  $X \times Y$ . Let  $\Phi_R$  and  $\Psi_R$  be as in Definition 1. Then  $\Phi_R$  and  $\Psi_R$  are homotopy equivalent.*

Every nonvoid simplicial complex  $\Sigma$  determines a partially ordered set  $\mathfrak{F}(\Sigma)$  called the *face poset* of  $\Sigma$ . The elements of this poset are the *nonempty* simplices of  $\Sigma$ , partially ordered by set inclusion. (Recall that 'poset' is short for 'partially ordered set'.)

For the finite setting, the homotopy equivalence of Dowker's Theorem may be seen by explicit formulas for maps between the face posets of the two Dowker complexes. These maps describe what is known as a *Galois Connection*. [This construction also appears as a core tool within the field of Formal Concept Analysis [21, 9].] Here are the formulas:

$$\begin{aligned}
 \phi_R &: \mathfrak{F}(\Psi_R) \rightarrow \mathfrak{F}(\Phi_R) & \psi_R &: \mathfrak{F}(\Phi_R) \rightarrow \mathfrak{F}(\Psi_R) \\
 \sigma &\mapsto \bigcap_{x \in \sigma} Y_x & \gamma &\mapsto \bigcap_{y \in \gamma} X_y
 \end{aligned}$$

These two maps are inverse homotopy equivalences. One sees this by considering the maps  $\phi_R \circ \psi_R$  and  $\psi_R \circ \phi_R$ . These compositions turn out to be what are called *closure operators* on the face posets  $\mathfrak{F}(\Phi_R)$  and  $\mathfrak{F}(\Psi_R)$ , respectively, implying that each is homotopic to the identity map, thereby establishing the desired homotopy equivalence. See Appendix B for detailed computations; see the next subsection for interpretation.

## 4.2 Inference from Closure Operators

A poset map  $f : P \rightarrow P$  is said to be a *closure operator* whenever  $x \leq f(x)$  and  $f(f(x)) = f(x)$  for all  $x \in P$ . If  $f$  is a closure operator, then it induces a homotopy equivalence between  $P$  and the image  $f(P)$  (see [1, 18]).

One can think of a closure operator as “pushing elements up” in the poset. From a privacy perspective, “pushing up” amounts to inference. Specifically,  $(\phi_R \circ \psi_R)(\gamma) \setminus \gamma$  consists of all additional attributes that may be inferred from observing attributes  $\gamma$ , while  $(\psi_R \circ \phi_R)(\sigma) \setminus \sigma$  consists of all additional individuals that may be inferred from observing individuals  $\sigma$ .

**Comment:** The formulas for  $\phi_R$  and  $\psi_R$  in Section 4.1 and the inference perspective extend to the empty simplex. Observe that  $\psi_R(\emptyset) = X$ , so  $(\phi_R \circ \psi_R)(\emptyset)$  consists of all attributes that every individual in  $X$  has. If  $(\phi_R \circ \psi_R)(\emptyset) \neq \emptyset$ , then the attributes  $(\phi_R \circ \psi_R)(\emptyset)$  are inferable “for free” from  $R$ , that is, without making any observations. Similarly,  $(\psi_R \circ \phi_R)(\emptyset)$  consists of all individuals who have every attribute in  $Y$ .

Any poset  $P$  defines a simplicial complex  $\Delta(P)$  called the *order complex* of  $P$ . The simplices of  $\Delta(P)$  are given by the finite chains  $\{p_0 < p_1 < \dots < p_n\}$  in  $P$ . Suppose we start with a simplicial complex  $\Sigma$ , construct its face poset  $\mathfrak{F}(\Sigma)$ , and then construct the order complex  $\Delta(\mathfrak{F}(\Sigma))$ . The result is isomorphic to the *first barycentric subdivision* of  $\Sigma$ . A convenient visualization of the face posets  $\mathfrak{F}(\Phi_R)$  and  $\mathfrak{F}(\Psi_R)$  therefore is to draw the barycentric subdivisions of  $\Phi_R$  and  $\Psi_R$ , respectively, as in Figure 4.

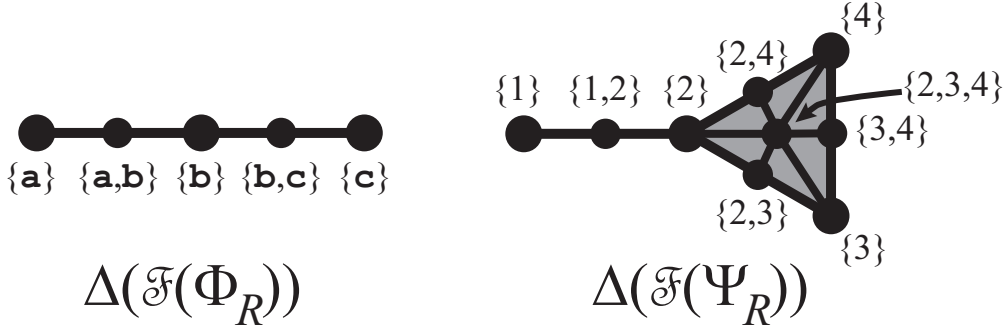


Figure 4: Order complexes of the face posets of the complexes  $\Phi_R$  and  $\Psi_R$  shown in Figure 3.

Viewed in the order complexes, functions  $\psi_R$  and  $\phi_R$  are easy to visualize. They are fully determined by their action on vertices of the order complexes, as shown in Table 1. (Bear in mind that each element of  $\mathfrak{F}(\Phi_R)$  represents a simplex in  $\Phi_R$  but is a vertex in  $\Delta(\mathfrak{F}(\Phi_R))$ . Similarly, each element of  $\mathfrak{F}(\Psi_R)$  represents a simplex in  $\Psi_R$  but is a vertex in  $\Delta(\mathfrak{F}(\Psi_R))$ .)

$\gamma$	$\psi_R(\gamma)$	$(\phi_R \circ \psi_R)(\gamma)$	$\sigma$	$\phi_R(\sigma)$	$(\psi_R \circ \phi_R)(\sigma)$
$\{a\}$	$\{1\}$	$\{a, b\}$	$\{1\}$	$\{a, b\}$	$\{1\}$
$\{b\}$	$\{1, 2\}$	$\{b\}$	$\{2\}$	$\{b, c\}$	$\{2\}$
$\{c\}$	$\{2, 3, 4\}$	$\{c\}$	$\{3\}$	$\{c\}$	$\{2, 3, 4\}$
$\{a, b\}$	$\{1\}$	$\{a, b\}$	$\{4\}$	$\{c\}$	$\{2, 3, 4\}$
$\{b, c\}$	$\{2\}$	$\{b, c\}$	$\{1, 2\}$	$\{b\}$	$\{1, 2\}$
			$\{2, 3\}$	$\{c\}$	$\{2, 3, 4\}$
			$\{3, 4\}$	$\{c\}$	$\{2, 3, 4\}$
			$\{2, 4\}$	$\{c\}$	$\{2, 3, 4\}$
			$\{2, 3, 4\}$	$\{c\}$	$\{2, 3, 4\}$

Table 1: The maps  $\psi_R$  and  $\phi_R$ , and their compositions, for relation  $R$  of Figure 3.

Using Table 1 one can again see how privacy loss might occur via  $R$ .

For instance, the map  $\phi_R \circ \psi_R$  gives rise to the closure (i.e., a “pushing up”)

$$\{a\} \xrightarrow{\psi_R} \{1\} \xrightarrow{\phi_R} \{a, b\},$$

telling us how to infer unobserved attribute **b** from observed attribute **a** (in the health study example of Section 3.1, Alice could infer that Bob HAS\_CANCER from knowing that he SMOKES).

Similarly, for the map  $\psi_R \circ \phi_R$ ,

$$\{4\} \xrightarrow{\phi_R} \{c\} \xrightarrow{\psi_R} \{2, 3, 4\},$$

leading to association inference (in the payroll example from Section 3.2, the payroll clerk could infer Bob and Mary's exposure to valuable information after learning of Julie's work on an important project).

Figure 5 indicates the homotopy deformations produced by the maps  $\phi_R \circ \psi_R$  and  $\psi_R \circ \phi_R$ , while Figure 6 show the resulting image of each face poset.

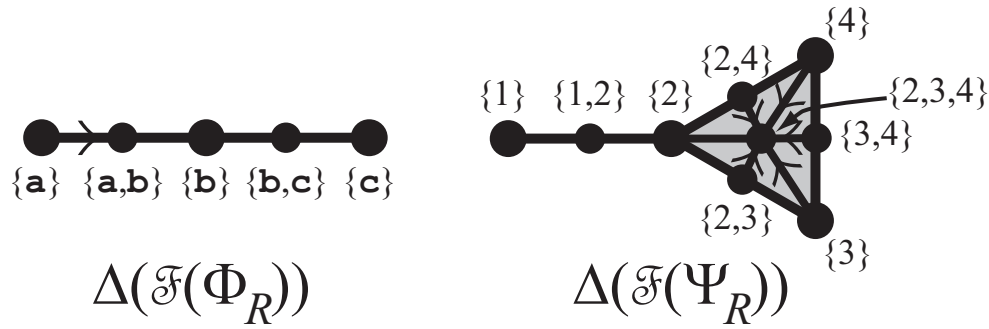


Figure 5: Closure operators  $\phi_R \circ \psi_R$  and  $\psi_R \circ \phi_R$  produce homotopy deformations, indicated by directed edges. In  $\mathfrak{F}(\Phi_R)$ ,  $\{a\}$  closes up to  $\{a, b\}$ . In  $\mathfrak{F}(\Psi_R)$ , most of the subsets of  $\{2, 3, 4\}$  close up to  $\{2, 3, 4\}$ . The exception is subset  $\{2\}$ , which does not move.

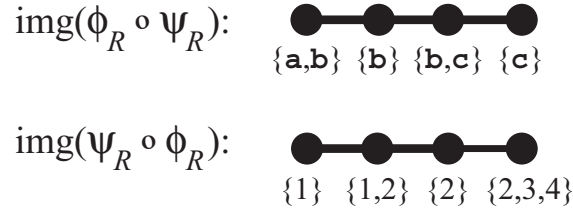


Figure 6: Result of the closure operators of Figure 5.

Observe that these two images are isomorphic. Matching up corresponding elements produces the poset  $P_R$  of Figure 1.

**Summary:** A relation  $R$  produces two simplicial complexes,  $\Phi_R$  and  $\Psi_R$ , one modeling attributes shared by individuals, the other modeling individuals with common attributes. The complexes are related by two maps,  $\phi_R$  and  $\psi_R$ , that are homotopy inverses. The compositions of these maps describe the attribute and association inferences possible via  $R$ , leveraging background information someone may have. These inferences are summarized by a poset  $P_R$  that pairs sets of individuals with sets of attributes. We may describe  $P_R$  as follows:

**Definition 3** (Doubly-Labeled Poset). *Let  $R$  be a relation on  $X \times Y$ .*

*The doubly-labeled poset  $P_R$  consists of all pairs of sets  $(\sigma, \gamma)$  such that  $\emptyset \neq \sigma \in \Psi_R$ ,  $\emptyset \neq \gamma \in \Phi_R$ ,  $\sigma = \psi_R(\gamma)$ , and  $\gamma = \phi_R(\sigma)$ .*

*The partial order on  $P_R$  is defined by:  $(\sigma_1, \gamma_1) \leq (\sigma_2, \gamma_2)$  if and only if  $\sigma_1 \subseteq \sigma_2$  (and/or, equivalently,  $\gamma_1 \supseteq \gamma_2$ ).*

(This definition agrees with our intuition that  $P_R$  is both the image  $(\psi_R \circ \phi_R)(\mathfrak{F}(\Psi_R))$  and the image  $(\phi_R \circ \psi_R)(\mathfrak{F}(\Phi_R))$ , by Appendix B.)

### 4.3 Attribute and Association Privacy

Here are formal definitions for the intuition developed via the previous examples:

**Definition 4** (Attribute Privacy). *A relation  $R$  preserves attribute privacy precisely when  $\phi_R \circ \psi_R$  is the identity operator on the poset  $\mathfrak{F}(\Phi_R) \cup \{\emptyset\}$ .*

**Definition 5** (Association Privacy). *A relation  $R$  preserves association privacy precisely when  $\psi_R \circ \phi_R$  is the identity operator on the poset  $\mathfrak{F}(\Psi_R) \cup \{\emptyset\}$ .*

**Comment:** For notational simplicity, we frequently say simply that  $\phi_R \circ \psi_R$  is the identity on  $\Phi_R$  and/or that  $\psi_R \circ \phi_R$  is the identity on  $\Psi_R$ .

### 4.4 Disinformation Example Re-Revisited

Recall the relation  $R'$  of Figure 2, which is relation  $R$  of Figure 1 but with an added entry of disinformation. Figure 7 displays the resulting Dowker complexes and the actions of the closure operators. Figure 8 flattens out the poset  $P_{R'}$  of Figure 2, so one sees its triangle structure and how it is the image of the Dowker complexes under the closure operators for  $R'$ .



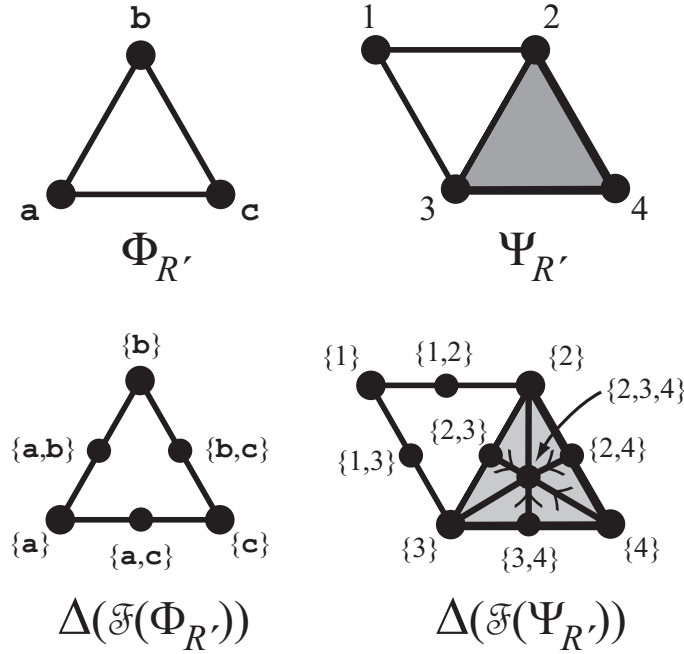


Figure 7: The Dowker complexes as well as the order complexes of their face posets for the relation  $R'$  of Figure 2. The closure operator  $\phi_{R'} \circ \psi_{R'}$  is the identity on  $\mathfrak{F}(\Phi_{R'}) \cup \{\emptyset\}$ . The closure operator  $\psi_{R'} \circ \phi_{R'}$  on  $\mathfrak{F}(\Psi_{R'}) \cup \{\emptyset\}$  closes many (but not all) subfaces of  $\{2, 3, 4\}$  up to  $\{2, 3, 4\}$ , as indicated by the directed arrows. The result is a poset isomorphic to the poset  $P_{R'}$  of Figure 2, drawn again slightly differently in Figure 8. Thus relation  $R'$  preserves attribute privacy but not association privacy.

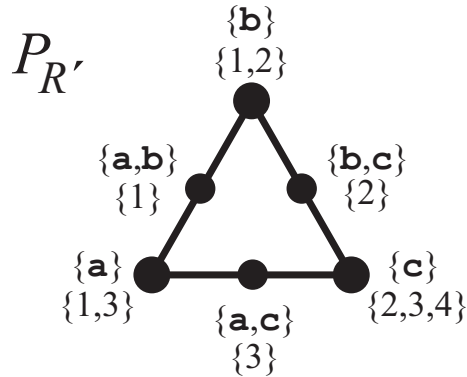


Figure 8: A flattened view of the doubly-labeled poset  $P_{R'}$  from Figure 2. Combined with Figure 7, this perspective shows how  $P_{R'}$  arises as the images of  $\mathfrak{F}(\Phi_{R'})$  and  $\mathfrak{F}(\Psi_{R'})$  under the closure operators  $\phi_{R'} \circ \psi_{R'}$  and  $\psi_{R'} \circ \phi_{R'}$ , respectively. (The vertices drawn as bigger dots in the current figure were higher up in the poset of Figure 2 than those drawn as smaller dots.)

## 5 The Face Shape of Privacy

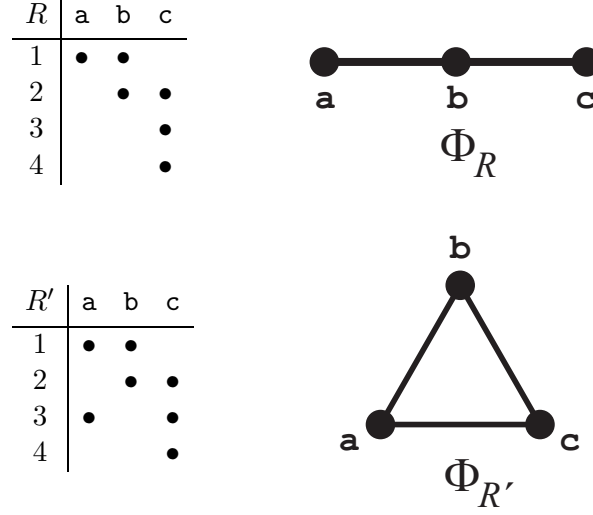


Figure 9: Relations  $R$  and  $R'$  of Section 3, along with their attribute complexes  $\Phi_R$  and  $\Phi_{R'}$ .

### 5.1 Free Faces

Figure 9 recapitulates relation  $R$  and  $R'$  from the previous two sections, along with their Dowker attribute complexes,  $\Phi_R$  and  $\Phi_{R'}$ , respectively. Recall that in  $R$  one could make the inference  $a \Rightarrow b$ , but no such inference was possible in  $R'$ .

Support for the inference  $a \Rightarrow b$  in  $R$  is evident in  $\Phi_R$ . No such support is evident in  $\Phi_{R'}$ . In particular, observe how vertex  $a$  has only one incident edge in  $\Phi_R$  but has two incident edges in  $\Phi_{R'}$ . The fact that there are two edges in  $\Phi_{R'}$ , with those edges being maximal simplices, means, intuitively, that vertex  $a$  is being “pulled” in two different inference directions, so one cannot conclude anything additional from  $a$ . In contrast, in  $\Phi_R$ ,  $a$  is being “pulled” only toward  $b$ , so it is possible that  $a$  implies  $b$ .

The underlying geometry is that of a free face. A simplex  $\sigma$  of a simplicial complex  $\Sigma$  is said to be a *free face* of  $\Sigma$  if it is a proper subset of exactly one maximal simplex of  $\Sigma$ . That is true for  $\{a\}$  in  $\Phi_R$  but not for  $\{a\}$  in  $\Phi_{R'}$ .

Of course, vertex  $\{c\}$  also forms a free face in  $\Phi_R$ , yet one cannot make any inferences upon observing just  $c$ . So, what is going on? The difference is that  $c$  is itself the only attribute of some individual in  $R$ . Even though  $\{c\}$  is technically a free face of  $\Phi_R$ , it is not really free to move under the closure operator  $\phi_R \circ \psi_R$ , whereas  $\{a\}$  is. Observe that individuals #2, #3, and #4 all have attribute  $c$ , but only individual #2 has additional attributes. This means that individuals #3 and #4 cannot ever be identified; they have effectively “camouflaged” themselves with individual #2.

If one disallows or disregards such camouflage, then the idea of a free face and privacy loss are equivalent. The following definition is useful:

**Definition 6** (Unique Identifiability). *Let  $R$  be a relation on  $X \times Y$  and suppose  $x \in X$ . We say that  $x$  is uniquely identifiable via relation  $R$  when  $\psi_R(Y_x) = \{x\}$ .*

Suppose  $R$  is a relation. Appendix C proves that if  $\Phi_R$  has no free faces, then  $R$  preserves attribute privacy. For the converse, Appendix C further proves that if  $R$  preserves attribute privacy *and* if every individual is uniquely identifiable, then  $\Phi_R$  has no free faces. (Dual statements hold for association privacy.)

## 5.2 Privacy versus Identifiability

Section 5.1 hinted at the difference between privacy and identifiability. In relation  $I$  below (“I” for “individuality” or “identity”), every individual has exactly one attribute that uniquely identifies that individual. Relation  $I$  *preserves privacy* fully. It is impossible to make any attribute inferences. If Bob reveals that he has attribute  $y_{\text{Bob}}$ , then Alice cannot infer any additional attributes for Bob. He has himself revealed everything about himself that there is to know, as far as relation  $I$  is concerned.

$I$	$y_1$	$y_2$	$\cdots$	$y_n$
$x_1$	•			
$x_2$		•		
$\vdots$			$\ddots$	
$x_n$				•

In contrast, all individuals in relation  $C$  (for “conformism”) have exactly the same set of attributes. As a result, there is *no privacy*: one can predict all the attributes of any individual in the relation without making any observations. On the other hand, no individual is identifiable.

$C$	$y_1$	$y_2$	$\cdots$	$y_n$
$x_1$	•	•	$\cdots$	•
$x_2$	•	•	$\cdots$	•
$\vdots$	$\vdots$	$\vdots$	$\ddots$	$\vdots$
$x_n$	•	•	$\cdots$	•

**Homogeneity:** Relation  $C$  exhibits a form of homogeneity often sought by anonymization or other privacy techniques. As we have suggested before, the utility of relation  $C$  is essentially zero, unless one makes the entries stochastic, so that some utility is encoded in the distribution.

The discussion of free faces in Section 5.1 suggests an alternative approach to homogeneity: one may preserve privacy and retain utility by choosing the geometry of the relation appropriately, for instance, so the space  $\Phi_R$  exhibits sphere-like homogeneity. There will be considerable discussion of the importance of spheres in the rest of the report.

## 5.3 Spheres and Privacy

The attribute complex  $\Phi_{R'}$  of Figure 9 is equal to a boundary complex, namely the boundary of the full simplex consisting of the attributes  $\{a, b, c\}$ . We will denote boundary complexes

by  $\partial(V)$ , with  $V$  some nonempty set. The simplices of  $\partial(V)$  are all proper subsets of  $V$ . Boundary complexes are homotopic to spheres, specifically  $\partial(V) \simeq \mathbb{S}^{n-2}$ , with  $n = |V|$ . For  $\Phi_{R'}$  of Figure 9, we have that  $\Phi_{R'} = \partial(\{\mathbf{a}, \mathbf{b}, \mathbf{c}\}) \simeq \mathbb{S}^1$ . (In English: The Dowker attribute complex is the boundary of a triangle, so homotopic to a circle.)

More generally, if for some relation  $R$ ,  $\Phi_R = \partial(Y)$ , then  $\Phi_R$  cannot have any free faces and so  $R$  preserves attribute privacy.

**Privacy and Utility:** An important observation is that boundary complexes exhibit homogeneity but still permit identifiability. If  $\Phi_R = \partial(Y)$  and no individual's attributes are a subset of another's attributes, then one can and needs to specify  $|Y| - 1$  attributes in order to identify an individual. The boundary structure ensures that one cannot infer any attributes by specifying fewer than  $|Y| - 1$  attributes, yet retains the ability to identify every individual.

Appendix J.1 gives an example of a contractible space that preserves attribute privacy. Observe, however, that the number of attributes needed to identify an individual in that example is considerably less than the total number of attributes in the space. For a boundary complex, it is just one less.

**Preserving Association and Attribute Privacy:** A consequence of these observations is that if one wishes to preserve both attribute and association privacy, then one requires both Dowker complexes to look like spheres. More specifically, either both Dowker complexes are linear cycles or both look like boundary complexes of the same dimension. In the latter case, the relation is isomorphic to a relation of the following form, in which the diagonal  $\{(x_i, y_i)\}$  is blank but all other entries are present:

$R$	$y_1$	$y_2$	$\cdots$	$\cdots$	$y_{n-1}$	$y_n$
$x_1$		•	•	$\cdots$	•	•
$x_2$	•		•	$\cdots$	•	•
$\vdots$	•	•		$\ddots$	$\vdots$	•
$\vdots$	$\vdots$	$\vdots$	$\ddots$		•	$\vdots$
$x_{n-1}$	•	•	$\cdots$	•		•
$x_n$	•	•	•	$\cdots$	•	

See Appendix E for further details.

## 5.4 A Spherical Non-Boundary Relation that Preserves Attribute Privacy

Consider relation  $R$  as in Figure 10. Relation  $R$  preserves attribute privacy, since  $\Phi_R$  has no free faces. The relation does not preserve association privacy. In particular, the quadrilaterals drawn for  $\Psi_R$  in the figure are actually tetrahedra. This means that the diagonals of the quadrilaterals are free faces. For instance, one would expect to infer individuals #1 and #6 as additional unobserved associates if one observes individuals #3 and #4. Indeed, computing using the closure operator  $\psi_R \circ \phi_R$ , we see that:

$$(\psi_R \circ \phi_R)(\{3, 4\}) = \psi_R(\{\mathbf{b}\}) = \{1, 3, 4, 6\}.$$

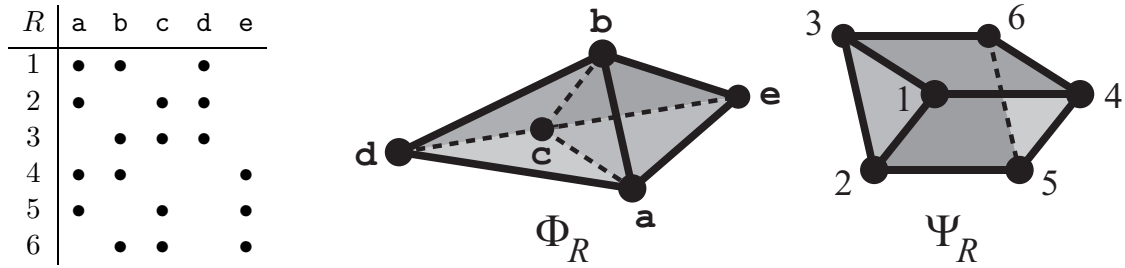


Figure 10: A relation  $R$  and its Dowker complexes  $\Phi_R$  and  $\Psi_R$ , each homotopic to the two-dimensional sphere  $\mathbb{S}^2$ . (One may view  $\Phi_R$  as two party hats glued together. One may view  $\Psi_R$  as a triangular cylinder with endcaps. However, the quadrilaterals drawn for the cylinder portion of  $\Psi_R$  are simply flattened sketches of what are actually solid tetrahedra.)

Relation  $R$  has another interesting feature. Even though  $\Phi_R$  is not itself a boundary complex, it is the simplicial join of two boundary complexes:

$$\Phi_R = \partial(\{a, b, c\}) * \partial(\{d, e\}).$$

In fact, we can think of  $R$  as  $R_1 \cup R_2$ , with  $R_1$  the restriction of  $R$  to the attributes  $\{a, b, c\}$  and  $R_2$  the restriction of  $R$  to the attributes  $\{d, e\}$ . This means that we can view every individual in  $R$  as being described by two *independent* attribute spaces. The attribute space  $\{d, e\}$  acts like a standard bit; every individual has exactly one of these two attributes. In contrast, the attribute space  $\{a, b, c\}$  is an “any 2 of 3” type of descriptor. Every individual has exactly two of these three attributes.

Figure 11 shows the relations  $R_1$  and  $R_2$  along with their Dowker attribute complexes.

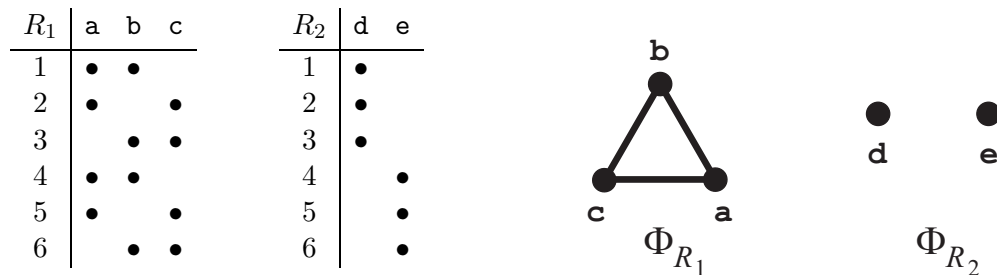


Figure 11: Relation  $R$  of Figure 10 decomposes into two disjoint relations  $R_1$  and  $R_2$  such that  $\Phi_R = \Phi_{R_1} * \Phi_{R_2}$ , with  $\Phi_{R_1}$  the boundary of a triangle and  $\Phi_{R_2}$  two isolated points. This means every individual in  $R$  has attributes that act like two independent coordinates: an “any 2 of 3” component and a bit.

## 6 Conditional Relations as Simplicial Links

The decomposition of Figures 10 and 11 is reminiscent of stochastic independence expressed as multiplication of probabilities. Similarly, there is a combinatorial analogue to the notion of a *conditional probability distribution*. It appears as the *link* of a simplex in a simplicial complex.

Given a relation  $R$ , suppose we have observed attributes  $\gamma$  for some unknown individual. The remaining possible combinations of attributes we might yet observe are described by the simplicial complex  $\text{Lk}(\Phi_R, \gamma) = \{\tau \in \Phi_R \mid \tau \cap \gamma = \emptyset \text{ and } \tau \cup \gamma \in \Phi_R\}$ . Interpretation:  $\tau \cap \gamma = \emptyset$  means that  $\tau$  consists of as yet unobserved attributes, while  $\tau \cup \gamma \in \Phi_R$  means that there is some individual who has the attributes  $\tau$  in addition to the attributes  $\gamma$  that have already been observed.

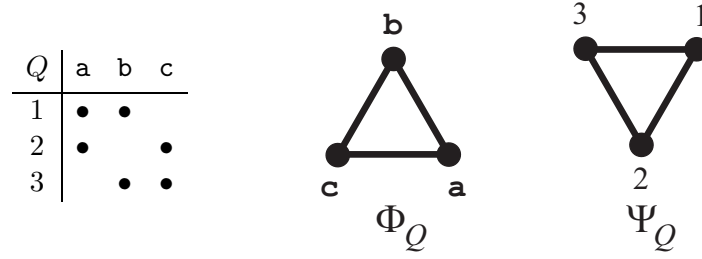


Figure 12: Relation  $Q$  describes the conditional relation resulting from  $R$  of Figure 10 upon observing attribute  $d$ . Note that  $\Phi_Q = \text{Lk}(\Phi_R, \{d\})$ .

For instance, after observing attribute  $d$  in relation  $R$  of Figure 10, we may conclude that we are observing one of the individuals in  $\{1, 2, 3\}$  and that the remaining attributes we might yet observe are any two attributes drawn from  $\{a, b, c\}$ . We can express these conclusions as yet another relation, namely the relation  $Q$  of Figure 12. Relation  $Q$  describes exactly which individuals could give rise to which attributes, consistent with the observation of  $d$  already made. **Thus  $\Phi_R$  plays a role much like a probability distribution, while  $\Phi_Q$  plays the role of a conditional distribution.** For another example, suppose we have observed attribute  $b$  in  $R$ . Then the resulting conditional relation  $Q'$  is as in Figure 13.

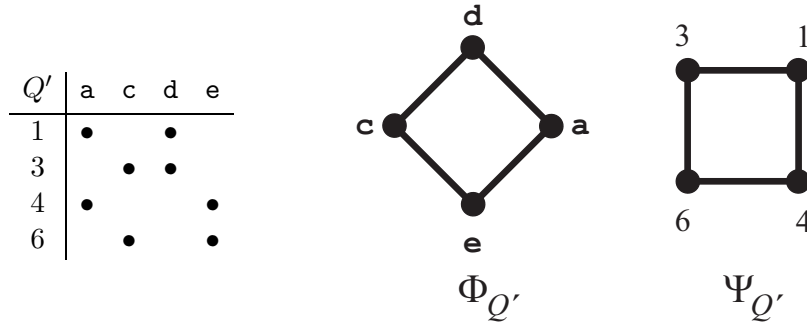


Figure 13: Relation  $Q'$  describes the conditional relation resulting from  $R$  of Figure 10 upon observing attribute  $b$ . Here  $\Phi_{Q'} = \text{Lk}(\Phi_R, \{b\})$ . The attribute space for  $Q'$  now factors into two independent bits:  $\{a, c\}$  constitutes one bit,  $\{d, e\}$  the other. This factoring is *conditional* on having observed  $b$ .

The formal constructions of conditional relations proceed as follows (a symbol of the form  $R|_W$  means “restrict  $R$  to  $W$ ”). See also Appendix C.

**Definition 7** (Conditional Attribute Relations). *Let  $R$  be a relation on  $X \times Y$  and suppose  $\gamma \subseteq Y$ . The following relation  $Q$  models  $\text{Lk}(\Phi_R, \gamma)$ :*

$$Q = R|_{\sigma \times \bar{Y}}, \quad \text{with } \sigma = \psi_R(\gamma) \quad \text{and} \quad \bar{Y} = \bigcup_{x \in \sigma} Y_x \setminus \gamma.$$

*The Dowker complexes are defined in the standard way, except for this special case: If  $\bar{Y} = \emptyset$  and  $\sigma \neq \emptyset$ , we let  $\Phi_Q$  and  $\Psi_Q$  be instances of the empty complex  $\{\emptyset\}$ .*

**Observe:**  $\text{Lk}(\Phi_R, \gamma) = \Phi_Q$  (a proof appears in Appendix C).

**Comment:** If  $\gamma \not\subseteq \Phi_R$ , then  $\sigma = \emptyset$  and  $Q$  is void, and so  $\Phi_Q$  is void, consistent with the standard definition of  $\text{Lk}(\Phi_R, \gamma)$  being void in this situation.

There is a dual construction for links of individuals  $\sigma$  in the Dowker complex modeling associations:

**Definition 8** (Conditional Association Relations). *Let  $R$  be a relation on  $X \times Y$  and suppose  $\sigma \subseteq X$ . The following relation  $Q$  models  $\text{Lk}(\Psi_R, \sigma)$ :*

$$Q = R|_{\bar{X} \times \gamma}, \quad \text{with } \gamma = \phi_R(\sigma) \quad \text{and} \quad \bar{X} = \bigcup_{y \in \gamma} X_y \setminus \sigma.$$

*The Dowker complexes are defined in the standard way, except for this special case: If  $\bar{X} = \emptyset$  and  $\gamma \neq \emptyset$ , we let  $\Psi_Q$  and  $\Phi_Q$  be instances of the empty complex  $\{\emptyset\}$ .*

**Observe:**  $\text{Lk}(\Psi_R, \sigma) = \Psi_Q$ .

As we will see in Section 7, the complex  $\text{Lk}(\Psi_R, \{x\})$  is useful for characterizing individual  $x$ ’s attribute privacy. If that seems surprising, observe that  $\text{Lk}(\Psi_R, \{x\})$  models the connections  $x$  has to other individuals. Those connections determine whether in  $\Phi_Q$ , and thus back in  $\Phi_R$ , there are attributes of  $x$  that are “free to move” under the closure operators.

## 7 Privacy Characterization via Boundary Complexes

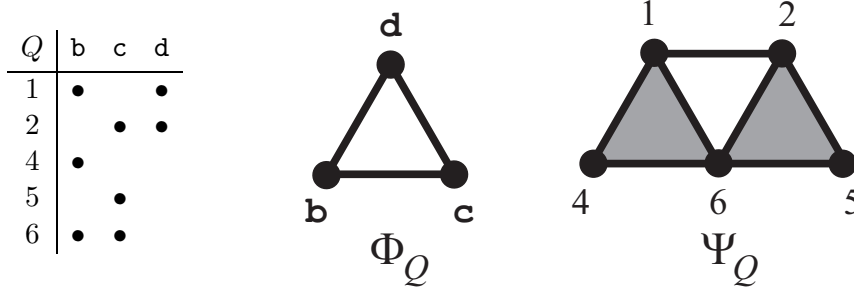


Figure 14: With  $R$  as in Figure 10, relation  $Q$  describes the conditional relation corresponding to  $\text{Lk}(\Psi_R, \{3\})$ . Also shown are the Dowker complexes of  $Q$ . By design,  $\Psi_Q = \text{Lk}(\Psi_R, \{3\})$ . Observe that  $\Phi_Q$  is the boundary complex  $\partial(\{b, c, d\})$ , with  $\{b, c, d\}$  being all of individual #3's attributes in relation  $R$ . That boundary condition characterizes full privacy for an individual.

We observed earlier that every individual in relation  $R$  of Figure 10 has full attribute privacy. We came to that conclusion after observing that  $\Phi_R$  has no free faces. In fact, one can focus in on the privacy of a single individual rather than look at the full relation. Let's pick one such individual, say #3, and look at the conditional relation  $Q$  that models the link  $\text{Lk}(\Psi_R, \{3\})$ , as shown in Figure 14.

Individual #3 has attributes  $\{b, c, d\}$  in  $R$ . The attribute complex  $\Phi_Q$  for  $Q$  is the boundary complex on exactly this set. Interpretation: for any nonempty proper subset of individual #3's attributes, some combination of *other* individuals in  $R$  has at least those attributes, but not all of individual #3's attributes. Moreover, there is a different combination of individuals for each proper subset that is missing exactly one of #3's attributes. That diversity of individuals ensures #3's attribute privacy.

The previous example suggests the following characterization: An individual has full attribute privacy precisely when the attribute complex of the individual's link is the boundary complex of the individual's attributes. Observe that this characterization is local to the individual; it does not depend on other individuals having privacy. We now formalize this intuition. Proofs appear in Appendix E.

Recall Definitions 4 and 6, from pages 20 and 22, respectively, formalizing the notions of privacy preservation and unique identifiability. And recall the semantics of  $P_R$ , for instance from Definition 3 on page 20.

**Theorem 9** (Individual Attribute Privacy). *Let  $R$  be a relation on  $X \times Y$ , with  $|X| > 1$ . Suppose  $x \in X$  is uniquely identifiable via  $R$ . Let  $Q$  be the relation modeling  $\text{Lk}(\Psi_R, x)$ . Then the following three conditions are equivalent:*

- (a)  $R$  preserves attribute privacy for  $x$ ,
- (b)  $\text{Lk}(\Psi_R, x) \simeq \mathbb{S}^{k-2}$ , with  $k = |Y_x|$ ,
- (c)  $\Phi_Q = \partial(Y_x)$ .



The previous theorem generalizes to sets of individuals for sets that are “stable” under the closure operators, i.e., that appear as the “set of individuals component” in an element of  $P_R$ :

**Theorem 10** (Group Attribute Privacy). *Let  $R$  be a relation on  $X \times Y$ . Suppose  $(\sigma, \gamma) \in P_R$ , with  $\sigma \neq X$ . Let  $Q$  be the relation modeling  $\text{Lk}(\Psi_R, \sigma)$ . Then the following three conditions are equivalent:*

- (a)  $(\phi_R \circ \psi_R)(\gamma') = \gamma'$ , for every subset  $\gamma'$  of  $\gamma$ ,
- (b)  $\text{Lk}(\Psi_R, \sigma) \simeq \mathbb{S}^{k-2}$ , with  $k = |\gamma|$ ,
- (c)  $\Phi_Q = \partial(\gamma)$ .

The following lemma appears in Appendix E:

**Lemma 11** (Interpreting Local Operators). *Let  $R$  be a relation on  $X \times Y$ .*

*Suppose  $(\sigma, \gamma) \in P_R$ , with  $\sigma \neq X$ .*

*Let  $Q$  be the relation on  $\overline{X} \times \gamma$  that models  $\text{Lk}(\Psi_R, \sigma)$  and suppose  $\overline{X} \neq \emptyset$ .*

*Then, for every  $\gamma' \subseteq \gamma$ :* (i) *If  $\gamma' \notin \Phi_Q$ , then  $\psi_R(\gamma') = \sigma$ ,*

*(ii) If  $\gamma' \in \Phi_Q$ , then  $\psi_R(\gamma') \supsetneq \sigma$ .*

*Moreover, in this case:*

*If  $(\phi_Q \circ \psi_Q)(\emptyset) = \emptyset$ , then  $(\phi_R \circ \psi_R)(\emptyset) = \emptyset$ .*

*If  $\gamma' \neq \emptyset$ , then  $(\phi_Q \circ \psi_Q)(\gamma') = (\phi_R \circ \psi_R)(\gamma')$ .*

The lemma says that observations of attributes that are consistent in  $Q$  have as interpretation more individuals in  $R$  than just the individuals  $\sigma$ , but if ever those observations become inconsistent in  $Q$ , then one has identified  $\sigma$  in  $R$ . Here “inconsistent in  $Q$ ” means that the observed attributes are legitimate attributes for  $Q$  but do not constitute a simplex of  $\Phi_Q$ . (Note: Such observed attributes necessarily constitute a simplex of  $\Phi_R$  since they are a subset of  $\gamma \in \Phi_R$ ).

Moreover, attribute inferences are identical in  $R$  and  $Q$  for nonempty simplices of  $\Phi_Q$ .

## 8 The Meaning of Holes in Relations

We have seen how spheres characterize privacy. More generally, when working with topological spaces, holes are significant. One wonders what holes mean for relations.

- Some holes arise as a consequence of exclusion between attributes, as we saw in the decomposition of Figures 10 and 11.

Sticking with binary exclusions, suppose a group of individuals are described by  $k$  bits. One can model those individuals via a relation containing  $2k$  attributes (one attribute for each possible bit value). Every individual has exactly  $k$  of those  $2k$  attributes. If all possible  $2^k$  combinations of bit values are represented by individuals in the relation, then the two Dowker complexes are both homotopic to  $\mathbb{S}^{k-1}$ , the sphere of dimension  $k-1$ . In fact,  $\Phi_R$  is the simplicial join of  $k$  copies of  $\mathbb{S}^0$ , while  $\Psi_R$  is visualizable as a hypercube in  $k$  dimensions, with  $(k-1)$ -dimensional subcubes fattened to be simplices. Figures 15, 16, and 17 depict the cases  $k = 1, 2$ , and  $3$ , respectively.

In short,  $k$  bits means a hole of dimension  $k-1$ , *if* all possible individuals are actually present in the relation.

(The lack of an expected hole may mean that the capacity of a relation has not been exhausted, hinting at possible inference. See Appendix J.3.)

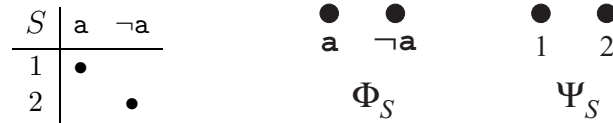


Figure 15: Relation  $S$  describes two individuals in terms of a single attribute and its negation. The topology of the Dowker complexes is  $\mathbb{S}^0$ .

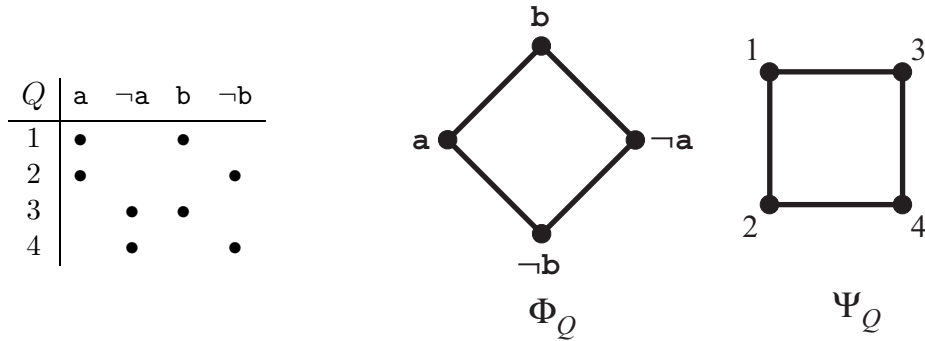


Figure 16: Relation  $Q$  describes four individuals in terms of two attributes and their negations. The topology of the Dowker complexes is  $\mathbb{S}^1$ .

- Minimal nonfaces (which may or may not be topological holes) suggest restrictions of a relation to equal-numbered attributes and individuals for whom there is both attribute

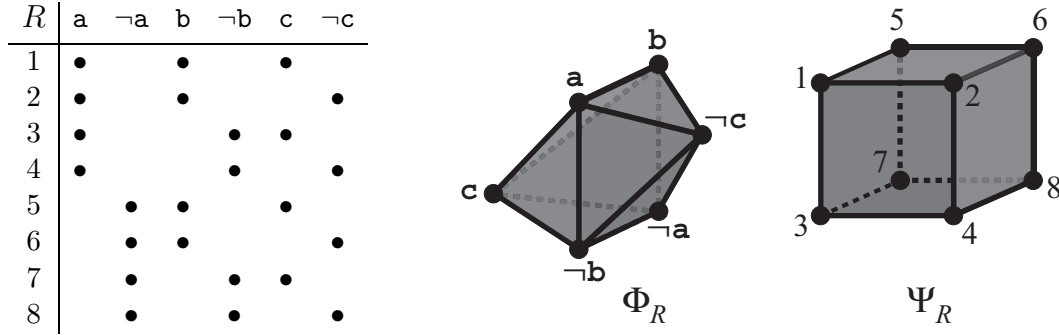


Figure 17: Relation  $R$  describes eight individuals in terms of three attributes and their negations. The topology of the Dowker complexes is  $\mathbb{S}^2$ . The cube faces are actually tetrahedra, flattened to parallelograms in the drawing.

and association privacy, within the restricted relation. This observation follows from the following results (here we assume that each relation has no blank rows or columns):

- A relation with more attributes than individuals cannot fully preserve attribute privacy.
- A relation with more individuals than attributes cannot fully preserve association privacy.
- A relation that preserves both attribute and association privacy must have the same number of attributes and individuals. Moreover, if the relation is connected, then both Dowker complexes are either linear cycles of the same length or they are both boundary complexes of full simplices, as we indicated previously.

See Appendices C and E for further details and proofs.

Consequently, minimal nonfaces of a relation may be viewed by restriction as descriptions of subrelations that preserve both attribute and association privacy.

- Minimal nonfaces can have other context-dependent meanings. For instance, in a certain authorship relation, knowing that each pair of three individuals has written a paper together appears to be a good predictor that all three individuals will co-author a paper together [13]. This suggests the following: if one sees that such an authorship hole does *not* fill over time, then one likely can infer some kind of obstruction, perhaps an incompatibility in the group as a whole, or the death of an author, for instance.
- When designing relations or anonymizing relations, these results suggest transformations that create “bubbly spaces” of some sort, in order to retain identifiability but also reduce unwanted inference. Sections 9 and J.2 discusses examples.
- Whatever holes there are in  $\Phi_R$  and  $\Psi_R$  must also show up in the poset  $P_R$ , since that poset is formed by homotopy equivalences from  $\Phi_R$  and  $\Psi_R$ . Interestingly, whereas one thinks of  $\Phi_R$  and  $\Psi_R$  simply as spaces, one sees a partial order on  $P_R$ . Something can

move, “up” or “down”. The elements of  $P_R$  are inference-stable, by design. So, what is this possible motion? It is a dynamic process that describes how information changes interpretation. For instance, as an individual reveals information about him- or herself, an observer can attempt to identify the individual, by finding interpretations in  $P_R$  of the information revealed. As the individual reveals additional information, the observer’s interpretation moves downward in  $P_R$ , narrowing the set of individuals.

Holes in the spaces  $\Phi_R$  and  $\Psi_R$  (and thus  $P_R$ ) constrain how that interpretation moves downward in  $P_R$ . The greater a hole’s dimension, the further a downward path has to move before identifying an individual. One can think of holes in a relation much like boulders in a stream. Eventually, the current of information sweeps past the hole, but it is forced to divert its motion, covering more distance. Moreover, there may be many paths around the hole, much like a leaf in a stream may divert around a boulder in different directions. The individual can force a particular path by choosing to reveal attributes in a particular order.

Much of the rest of the report explores the implications of this stream analogy. The analogy merges with the realization that privacy is a dynamic process, certain to flow toward identification when attributes are static or persistent, yet subject to channeling and turbulence when fluid. See in particular Section 10 onward.

## 9 Change-of-Attribute Transformations

Free faces and holes in the Dowker complex  $\Phi_R$  can sometimes suggest changes in attributes that preserve desired information but reduce inference. Consider the “ice-cream cone” relation  $C$  of Figure 18 and the corresponding complexes shown in Figure 19. The relation describes four individuals in terms of the two-flavor two-scoop ice-cream cones each individual enjoys at a particular ice-cream parlor.

$C$	gc	gs	cs	cv	sv	gv
Bob	•	•	•			
Alice			•	•	•	
David		•			•	•
Cindy	•			•		•

g = ginger  
c = chocolate  
s = strawberry  
v = vanilla

Figure 18: Four individuals and their preferences for ice-cream cones containing two scoops, with different flavors (each letter represents a flavor, as indicated). See Figure 19 for the Dowker complexes.

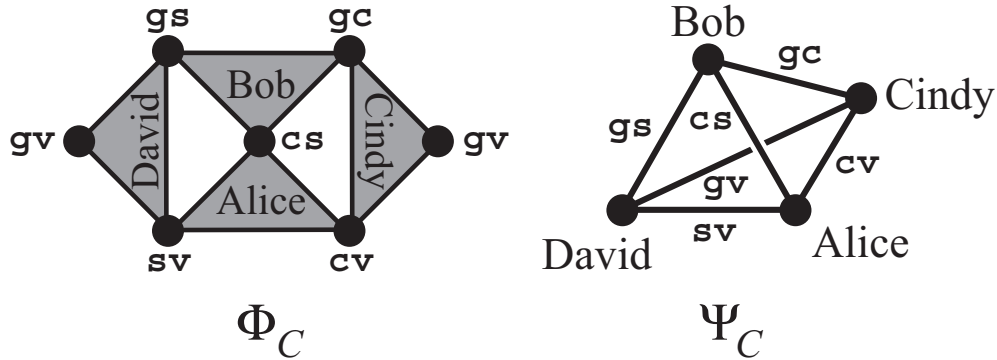


Figure 19: The Dowker complexes for the relation of Figure 18.  $\Phi_C$  is a complex whose vertices are ice-cream cones (two flavors). (For visualization purposes, the complex is flattened, with the leftmost and rightmost vertices really representing the same ice-cream cone.) Each maximal simplex is a triangle, labeled with the individual who enjoys the three types of cones comprising the triangle.  $\Psi_C$  is a complex whose vertices are individuals. Each maximal simplex is an edge, representing a two-flavor two-scoop ice-cream cone that each of two individuals enjoys; the edge is labeled with the cone flavors. The homotopy type of each complex is  $\mathbb{S}^1 \vee \mathbb{S}^1 \vee \mathbb{S}^1$ .

Relation  $C$  is a typical “2-implies-3” relation: Any two different ice-cream cones fully identify an individual, thereby implying a third ice-cream cone, as can be seen from either Dowker complex: In  $\Phi_C$ , every edge is a free face of its encompassing triangle. Moreover, the edge is not itself generated by any individual.<sup>2</sup> The closure operator  $\phi_C \circ \psi_C$  must therefore map every edge to a triangle. Similarly, in  $\Psi_C$ , any two edges intersecting at a vertex imply the third edge incident on that vertex.

<sup>2</sup>We say an individual  $x$  of a relation  $R$  generates a simplex  $\gamma \in \Phi_R$  when  $\gamma = Y_x$ . Similarly, an attribute  $y$  generates a simplex  $\sigma \in \Psi_R$  when  $\sigma = X_y$ .

This type of relation models, in the small, inferences such as those reported in [17, 15]. For instance, [17] reported that zip code, gender, and birth date were likely sufficient in 1990 to identify 87% of individuals in the U.S. That is nearly a “3-implies-all” type of relation. Similarly, [15] reported that 8 movie ratings and dates were enough to uniquely identify 99% of viewers in the Netflix Prize dataset. That is essentially an “8-implies-all” type of relation.

Let’s focus for a moment on Bob’s neighborhood. That relation, let’s call it  $B$ , and its complexes are depicted in Figure 20. (The relation models  $\overline{\text{St}}(\Psi_C, \{\text{Bob}\})$ , see Appendix C.)

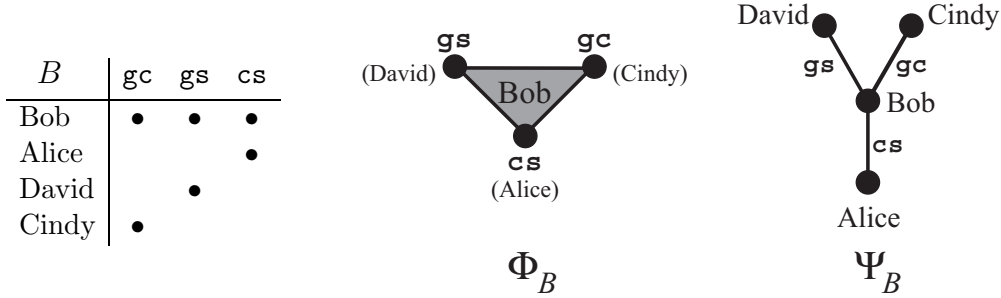


Figure 20: Relation  $B$  models Bob’s neighborhood in the ice-cream relation of Figure 18. Each maximal simplex is labeled with its generator. Generators of non-maximal simplices are indicated in parentheses.

As in  $C$ , seeing someone eat one ice-cream cone is not enough to identify anyone in  $B$ . Seeing someone (in this case Bob), eat two *different* types of ice-cream cones, is sufficient to infer the third type of ice-cream cone that individual prefers. How might we prevent this? We observe that the vertices of  $\Phi_B$  are themselves generated by individuals while the edges are not. Homotopically, therefore, we want to expand the vertices of  $\Phi_B$  into edges, and contract the edges of  $\Phi_B$  into vertices. One possible way to accomplish this is the take logical ORs of the existing attributes. With  $\oplus$  meaning Boolean OR, we define:

$$\alpha = gc \oplus gs, \quad \beta = gc \oplus cs, \quad \gamma = gs \oplus cs.$$

Then relation  $B$  becomes  $B'$  as in Figure 21. The result is that the free faces of  $\Phi_{B'}$  now are generated by other individuals, so even though they are free, the closure operator does not move them. In fact, the closure operator  $\phi_{B'} \circ \psi_{B'}$  is the identity on  $\mathfrak{F}(\Phi_{B'}) \cup \{\emptyset\}$ , meaning that no attribute inference is possible in  $B'$ .

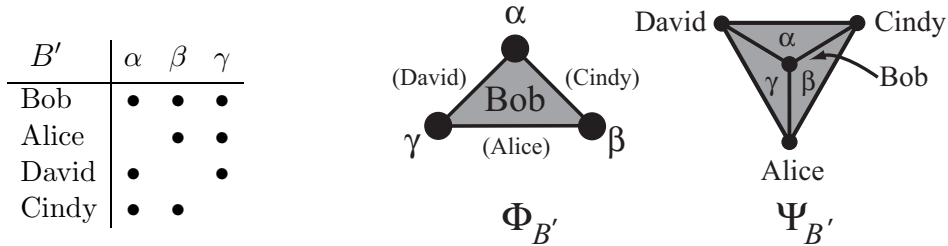


Figure 21: Relation  $B'$  represents relation  $B$  of Figure 20, now with a coordinate transformation for the attributes. Simplices are again labeled by generators.

Now imagine performing similar operations for all four individuals of relation  $C$  from Figure 18. One winds up constructing four logical ORs:

$$gc \oplus gs \oplus gv, \quad gc \oplus cs \oplus cv, \quad gs \oplus cs \oplus sv, \quad cv \oplus sv \oplus gv.$$

Two observations:

1. Each OR describes three ice-cream cones that form a hole in the complex  $\Phi_C$  of Fig. 19.
2. Each such hole may be interpreted as a single flavor, namely the flavor in common to the three ice-cream cones appearing in the OR. For instance, “ginger” (abbreviated as  $g$ ) is the common flavor for the OR  $gc \oplus gs \oplus gv$ .

In order to describe the resulting relation, it is perhaps easiest to express those four new coordinates themselves via a relation  $S$  that describes the scoops present in an ice-cream cone:

$S$	$g$	$c$	$s$	$v$
$gc$	•	•		
$gs$	•		•	
$cs$		•	•	
$cv$		•		•
$sv$			•	•
$gv$	•			•

Finally, to perform the coordinate-transformation, one simply multiplies Boolean matrices, with addition being Boolean OR and multiplication being Boolean AND:  $F = CS$ . The relation  $F$  and its complexes appear in Figure 22.

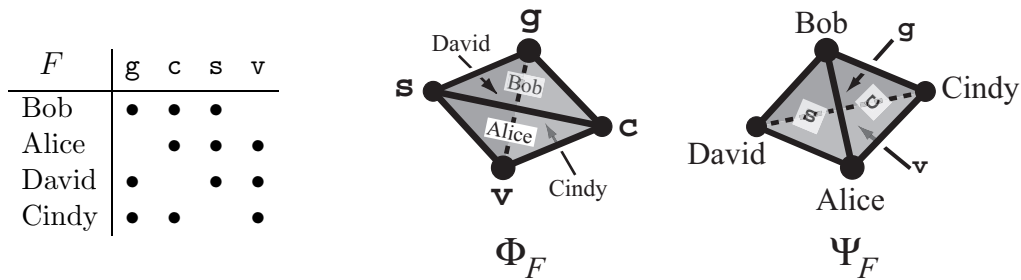


Figure 22: Relation  $F$  describes the individual flavors each individual prefers.  $\Phi_F$  is the boundary of a tetrahedron, with flavors as vertices.  $\Psi_F$  is the Dowker dual of  $\Phi_F$ , meaning it too is the boundary of a tetrahedron, now with the roles of flavors and individuals interchanged. For both  $\Phi_F$  and  $\Psi_F$ , each maximal simplex is a triangle, labeled with its generator.

Relation  $F$  represents a description of the four individuals' preferences in terms of flavors not cones. The resulting complexes  $\Phi_F$  and  $\Psi_F$  are now boundary complexes of full simplices, each homeomorphic to  $\mathbb{S}^2$ . These complexes have no free faces, so no inference is possible. Observe further that  $\Phi_F$  is homotopic to what one obtains from  $\Phi_C$  by filling the  $\mathbb{S}^1$ -holes.

Indeed, this idea implicitly motivated our construction, as a way to remove free faces. Similarly,  $\Psi_F$  is isomorphic to what one obtains from  $\Psi_C$  by filling its  $\mathbb{S}^1$ -holes.

One should ask how this approach might generalize. The answer is mixed. The idea of removing free faces is central. There are many ways to accomplish that, with relational composition being but one method. One issue with logical ORs is that it is very easy to obtain an OR that is always TRUE, at which point the resulting attribute is of little use.

Even with more general transformations, there remains the issue of whether the new attributes are grounded in what is actually measurable. In the ice-cream example, it was fortunate that cones decomposed naturally into flavors. It is at least plausible that someone might merely observe the flavors a customer prefers, not the combinations of flavors as cones. If, however, only cones can be observed, then one is forced to deal with relation  $C$  as given.



## 10 Leveraging Lattices for Privacy Preservation

This section examines more carefully the poset  $P_R$  along with its lattice structure, leading to the idea of *informative attribute release sequences*. These are attributes that an individual can release in a particular order so as to prevent inference of any attributes yet to be released via the sequence. The lattice length therefore describes the extent to which an individual can defer identification. Homology in  $P_R$  provides lower bounds on that length.

### 10.1 Attribute Release Order

Relation  $G$  of Figure 23 describes hypothetical co-authorships among five authors in producing travel guides for five European cities. Each collaboration consists of three authors working together on one of the five travel guides.

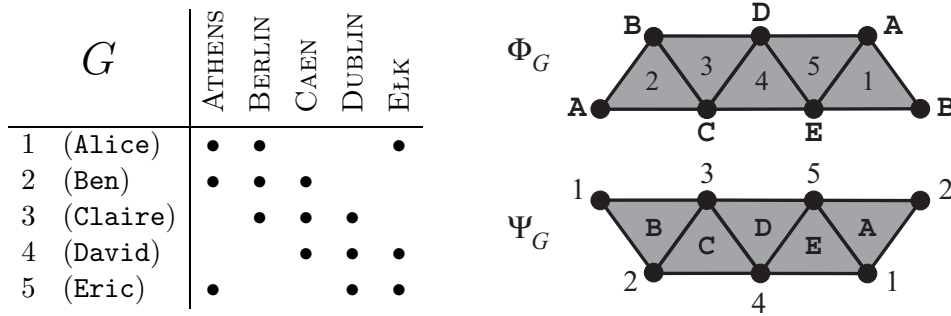


Figure 23: A relation  $G$  describing co-authorship of travel guides. The Dowker complexes are dual triangulations of the Möbius strip, with  $S^1$  homotopy type. (Notes: Integers indicate authors, letters indicate cities via first letter abbreviations. Some vertices and edges appear twice for ease of viewing. Each maximal simplex is labeled with its generating author or city.)

Suppose in casual conversation a person mentions that he/she worked on producing a travel guide for BERLIN. In the context of relation  $G$ , that information means the author is one of  $\{\text{Alice}, \text{Ben}, \text{Claire}\}$ . If the author further mentions working on the travel guide for DUBLIN, then that identifies the author as **Claire**. Equivalently, the listener can infer that the author also helped write the travel guide for CAEN. (This form of inference was the source of problems for the Netflix Prize [15].)

**Claire** was a co-author on three travel guides, for BERLIN, CAEN, and DUBLIN. Now consider the different possible sequential ways in which **Claire** might reveal which books she helped co-author, along with the point at which her identity becomes known (see Figure 24).

Of the six possible ways, four do not fully identify **Claire** until she has revealed all three books that she co-authored. However, two of the possible six release sequences do allow a listener to identify the author and infer an additional book that she co-authored.

This example shows how inference may be a dynamic process. While a consumer of data may wish to identify **Claire** with as little information as possible, the author herself may wish to delay that identification for as long as possible (perhaps for reasons of public mystery in selling books). In the example, the *minimal length* of an *identifying attribute release sequence*

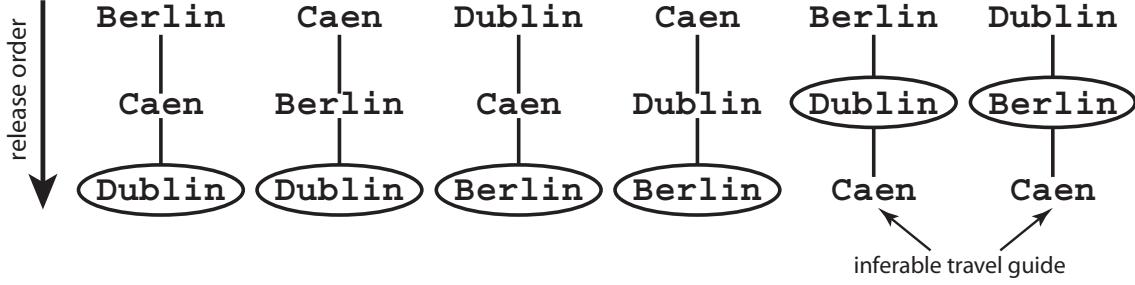


Figure 24: This figure shows the six possible sequential ways in which author #3 (Claire) of Figure 23 can mention the cities for which she co-authored travel guides. The point at which her identity becomes known in any such release sequence is circled. If **Claire** does not mention CAEN, one can infer, via relation  $G$  of Figure 23, that she co-authored a travel guide for that city as soon as she mentions the other two cities, BERLIN and DUBLIN, in either order.

is two, while the *maximal length* is three. If **Claire** can control how information is released, then she can choose to reveal what might otherwise be inferred, namely that she co-authored a travel guide to CAEN, thereby delaying her identification.

Finally, we observe that the order of attributes released may or may not matter. In the travel guide example, **Claire** should mention CAEN before the end of her revelations (if she wants to delay her identification), but the order of cities mentioned is otherwise irrelevant. The topology of the doubly-labeled poset  $P_G$  encodes this order (in)dependence, as we will see shortly. Indeed, much of the remainder of this report examines the connection between the topology of a relation’s doubly-labeled poset and the length of attribute release sequences.

## 10.2 Inferences on a Lattice

The doubly-labeled poset of a relation produces a lattice [21], as follows:

**Definition 12** (Galois Lattice). *Let  $R$  be a relation on  $X \times Y$ , with  $X \neq \emptyset$  and  $Y \neq \emptyset$ . Let  $P_R$  be the associated doubly-labeled poset.*

*(Recall from Definition 3 on page 20 that an element of  $P_R$  is a pair  $(\sigma, \gamma)$ , with*

$$\emptyset \neq \sigma = \psi_R(\gamma) \in \Psi_R \text{ and } \emptyset \neq \gamma = \phi_R(\sigma) \in \Phi_R.$$

*We previously defined a partial order on  $P_R$  by  $(\sigma_1, \gamma_1) \leq (\sigma_2, \gamma_2)$  iff  $\sigma_1 \subseteq \sigma_2$  (iff  $\gamma_1 \supseteq \gamma_2$ ).)*

*$P_R$  may already contain a bottom element of the form  $(\sigma, Y)$ , with  $\sigma$  those individuals in  $X$  who have all the attributes in  $Y$ . If not, we adjoin  $(\emptyset, Y)$  to the bottom of  $P_R$ .*

*$P_R$  may already contain a top element of the form  $(X, \gamma)$ , with  $\gamma$  those attributes in  $Y$  that every individual in  $X$  has. If not, we adjoin  $(X, \emptyset)$  to the top of  $P_R$ .*

*We refer to the resulting poset as the Galois lattice  $P_R^+$ . It has lattice operations  $\vee$  and  $\wedge$ :*

$$\begin{aligned} (\sigma_1, \gamma_1) \vee (\sigma_2, \gamma_2) &= ((\psi_R \circ \phi_R)(\sigma_1 \cup \sigma_2), \gamma_1 \cap \gamma_2), \\ (\sigma_1, \gamma_1) \wedge (\sigma_2, \gamma_2) &= (\sigma_1 \cap \sigma_2, (\phi_R \circ \psi_R)(\gamma_1 \cup \gamma_2)). \end{aligned}$$

*We sometimes refer to the bottom element of  $P_R^+$  by  $\hat{0}_R$  and to the top element by  $\hat{1}_R$ .*

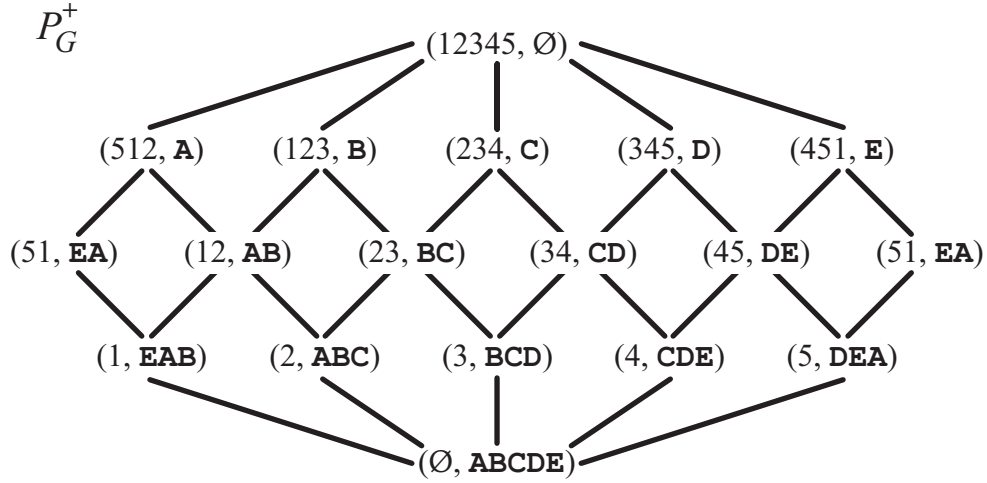


Figure 25: The lattice  $P_G^+$  for the travel guide relation of Figure 23. Each element is a pair of sets  $(\sigma, \gamma)$  such that  $\sigma = \psi_G(\gamma)$  and  $\gamma = \phi_G(\sigma)$ . (We have elided commas and braces in sets, for ease of viewing.) The lattice operations model inferences possible from observations. For instance,  $(123, B) \wedge (345, D) = (3, BCD)$ , meaning that observation of attributes B and D permits inference of additional attribute C and identification of author #3. (In Figure 23, attribute C is the travel guide for CAEN and author #3 is **Claire**.) The lattice wraps around, with element  $(51, EA)$  duplicated for ease of viewing. If one removes the top and bottom elements, the remaining poset  $P_G$  has  $\mathbb{S}^1$  homotopy type, just like the Möbius strip.

Figure 25 shows the lattice  $P_G^+$  for the travel guide relation of Figure 23. Observe how the lattice encodes attribute and association inferences (or lack thereof) via its lattice operations.

**Special Cases:** It can happen that the lattice consists of a single element. For example, with relation  $C$  as on page 23,  $P_C^+ = P_C = \{(X, Y)\}$ . In particular,  $\hat{0}_C = \hat{1}_C$ .

Definition 12 ignores the situation in which  $R$  is void. One possibility is to view  $P_R^+$  as void and  $P_R$  as degenerate.

### 10.3 Preserving Attribute Privacy for Sets of Individuals

Theorem 9 on page 28 described the conditions under which an individual has full attribute privacy. For such an individual, the order in which that individual (or anyone) releases the individual's attributes is irrelevant. Any order is fine. Only once all attributes have been released, can an observer definitively identify the individual. Theorem 10 described a similar result for certain sets of individuals, including those with whom an individual is confusable after only some of his/her attributes have been released.

Consider  $\text{Lk}(\Psi_G, 3)$ , modeled by relation  $C$  as in Figure 26. This relation describes the authors with whom **Claire** has collaborated, via their co-authored books. The Dowker complexes are contractible, so by either Theorem 9 or Theorem 10, we know that some attribute inference is possible involving **Claire**. Lemma 11 on page 29 tells us to look for a subset of  $\{\text{BERLIN}, \text{CAEN}, \text{DUBLIN}\}$  that is a simplex of  $\Phi_G$  but not of  $\Phi_C$ . As is apparent from the figure,

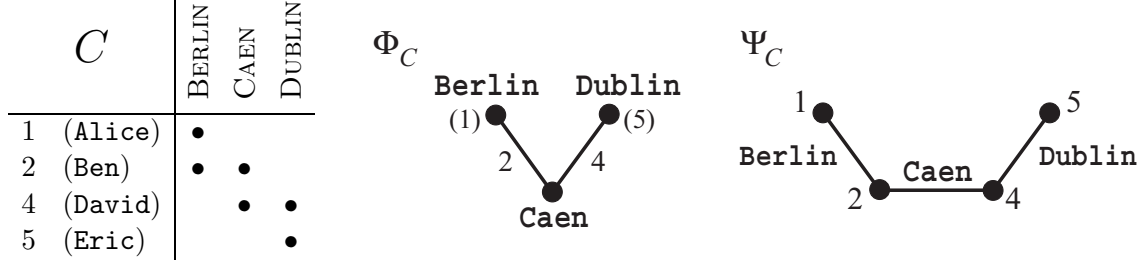


Figure 26: Relation  $C$  describes  $\text{Lk}(\Psi_G, 3)$ , the link of **Claire** in the relation of Figure 23. (Each maximal simplex in any one complex is labeled with its generating attribute or individual from the other complex. Generators of non-maximal simplices are indicated in parentheses.)

the set  $\{\text{BERLIN}, \text{DUBLIN}\}$  satisfies these conditions, consistent with our earlier observations. Alternatively, looking at  $P_C^+$  in Figure 27, we see that  $(12, \text{B}) \wedge (45, \text{D}) = (\emptyset, \text{BD})$ , allowing us to draw the same conclusion. Consequently, **Claire** should be sure to mention her travel guide for CAEN early on, not leave it for last, if she wants to delay identification.

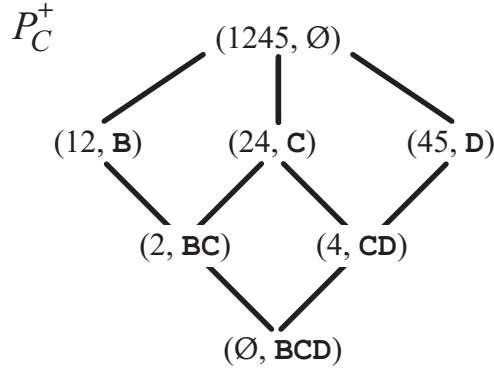


Figure 27: The lattice  $P_C^+$  for the link of **Claire**, as given in Figure 26. (Here authors appear as integer indices and city names appear as first letter abbreviations.) Observe that this lattice may be viewed as a sublattice of  $P_G^+$ , containing all elements that include individual #3 there, but with that individual removed here.

Now let us take this reasoning one step further. Consider an element of  $P_G^+$  corresponding to some state just prior to identification of **Claire**, for instance  $(23, \text{BC})$ . This element corresponds to both of the first two release sequences of Figure 24: **Claire** has mentioned her work regarding the travel guides for BERLIN and CAEN, but has not yet mentioned DUBLIN. Thus there is still some ambiguity as to her identity (it is either author #2 or author #3). In terms of Theorem 10 on page 29,  $\sigma = \{2, 3\}$ ,  $\gamma = \{\text{BERLIN}, \text{CAEN}\}$ , and  $k = 2$ .

Figure 28 shows the relation describing  $\text{Lk}(\Psi_G, \{2, 3\})$ . The Dowker complexes have  $\mathbb{S}^0$  homotopy type, thus satisfying the topological conditions of Theorem 10. Consequently, there is no attribute inference possible in the encompassing relation based on attributes that appear in the link. That means the order in which **Claire** releases the two attributes BERLIN and CAEN is immaterial. This conclusion is consistent with the conclusion one draws upon explicitly enumerating all release sequences, as in Figure 24.

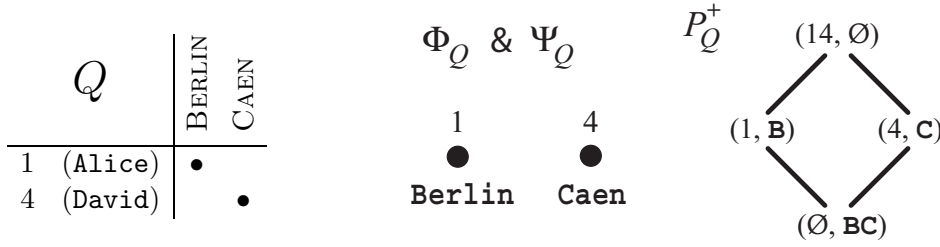


Figure 28: Relation  $Q$  describes  $\text{Lk}(\Psi_G, \{2, 3\})$ , the combined link of authors #2 and #3 (Ben and Claire) in the relation of Figure 23. These two authors have together collaborated with each of authors #1 and #4 (Alice and David) but have not both together collaborated with author #5 (Eric). The two Dowker complexes are each instances of  $\mathbb{S}^0$ , so essentially the same. The corresponding lattice  $P_Q^+$  is also very simple.

#### 10.4 Informative Attribute Release Sequences

This subsection defines more precisely the idea of controlled information release. These definitions will help us better understand holes in a relation's Dowker complexes, via Theorem 10. Subsequently, Section 11 will test that insight with data from the world wide web.

**Definition 13** (Attribute Release Sequence). *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. An attribute release sequence for  $R$  is a nonempty set of attributes from  $Y$  released in a particular sequential order:*

$$y_1, y_2, \dots, y_k, \quad \text{with } k \geq 1.$$

*We say that the sequence has length  $k$ .*

*We say that an attribute release sequence is informative if*

$$y_i \notin (\phi_R \circ \psi_R)(\{y_1, \dots, y_{i-1}\}), \quad \text{for all } 1 \leq i \leq k.$$

*(Note: for  $i = 1$ , the requirement states that  $y_1 \notin (\phi_R \circ \psi_R)(\emptyset) = \phi_R(X)$ .)*

*(We sometimes use the abbreviation 'iars' to mean either 'informative attribute release sequence' or 'informative attribute release sequences'.)*

**Interpretation:** When  $i = 1$ , the argument to  $\phi_R \circ \psi_R$  is the empty set, so the condition requires that  $y_1 \notin \phi_R(X)$ . In other words,  $y_1$  may not be any attribute that is shared by all individuals in  $X$ . Any such attribute could be inferred “for free” and thus would not be informative. Thereafter, the condition requires that any attribute to be released not be inferable from those already released.

We are interested in understanding the extent to which order of release matters:

**Definition 14** (Isotropy). *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. Suppose  $\emptyset \neq \gamma \subseteq Y$ .*

*We say that  $\gamma$  is isotropic if every possible ordering of all the elements in  $\gamma$  forms an informative attribute release sequence for  $R$ .*

We are interested in the minimal and maximal lengths of informative attribute release sequences:

**Definition 15** (Identification and Minimal Identification). *Let  $R$  be a relation on  $X \times Y$ .*

*We say that a set of attributes  $\gamma \subseteq Y$  identifies a set of individuals  $\sigma \subseteq X$  in  $R$  when  $\psi_R(\gamma) = \sigma$ . (We sometimes alternatively say that  $\gamma$  localizes (to)  $\sigma$ .)*

*We say that  $\gamma$  is minimally identifying (for  $\sigma$ ) if both the following conditions hold:*

- (i)  $\psi_R(\gamma) = \sigma$ ,
- (ii)  $\psi_R(\gamma') \supsetneq \sigma$  for every  $\gamma' \subsetneq \gamma$ .

**Definition 16** (Identification Lengths). *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. Suppose  $(\sigma, \gamma) \in P_R$ . Define the fast and slow attribute release lengths for  $\sigma$  as*

$$\begin{aligned} r_{\text{fast}}(\sigma) &= \min \{ |\gamma| \mid \gamma \in \Phi_R \text{ and } \psi_R(\gamma) = \sigma \}. \\ r_{\text{slow}}(\sigma) &= \max \{ k \mid y_1, \dots, y_k \text{ is an iars for } R \text{ and } \psi_R(\{y_1, \dots, y_k\}) = \sigma \}. \end{aligned}$$

An argument similar to that in Appendix D shows that the following problem is *NP*-complete: Given  $\sigma$ , is there some minimally identifying  $\gamma$  for  $\sigma$  of size at most  $k$ ?

## 10.5 Isotropy, Minimal Identification, and Spheres

There is no requirement in Definition 13 that an informative attribute release sequence be a simplex in  $\Phi_R$ . Indeed, when working with links, it is useful to create informative attribute release sequences that are not simplices in the link, thereby identifying a set of individuals in the encompassing relation, as per Lemma 11. However, it is always the case that any inconsistency arises only with the last attribute released:

**Lemma 17** (Almost a Simplex). *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty.*

*Suppose  $\{y_1, \dots, y_k\}$  is an informative attribute release sequence for  $R$ .*

*Then  $\{y_1, \dots, y_{k-1}\} \in \Phi_R$ .*

**Proof:** If  $\{y_1, \dots, y_{k-1}\} \notin \Phi_R$ , then  $(\phi_R \circ \psi_R)(\{y_1, \dots, y_{k-1}\}) = \phi_R(\emptyset) = Y$ . Since  $y_k \in Y$ , this contradicts the requirement of Definition 13.  $\square$

Interestingly, when an informative attribute release sequence is a simplex, then being isotropic is equivalent to being minimally identifying. Moreover, topologically, we can characterize this isotropy property as a sphere appearing via a restricted link:

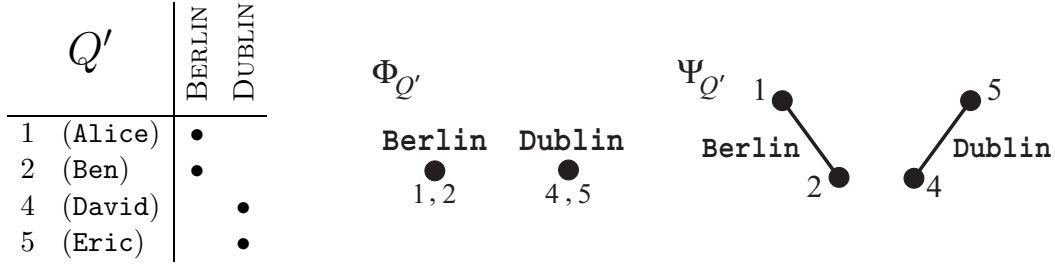


Figure 29: Relation  $Q' = Q(\sigma, \gamma)$ , for the book authorship example of Figure 23, with  $\sigma = \{3\}$  and  $\gamma = \{\text{BERLIN}, \text{DUBLIN}\}$ . Relation  $Q'$  describes the link of author #3 (Claire) restricted to the attribute set  $\{\text{BERLIN}, \text{DUBLIN}\}$ . See Figure 26 for the full link relation. (Each maximal simplex in the Dowker complexes is again labeled with its generating individuals or attribute.)

**Definition 18** (Restricted Link). *Let  $R$  be a relation on  $X \times Y$ .*

*Suppose  $\sigma \in \Psi_R$  and  $\gamma \subseteq \phi_R(\sigma)$ .*

*Define relation  $Q(\sigma, \gamma)$  as follows:*

$$Q(\sigma, \gamma) = R|_{\overline{X} \times \gamma}, \quad \text{with} \quad \overline{X} = \bigcup_{y \in \gamma} X_y \setminus \sigma.$$

*The Dowker complexes are defined in the standard way, except for these special cases:*

*If  $\sigma = X$ , we let  $\Psi_{Q(\sigma, \gamma)}$  and  $\Phi_{Q(\sigma, \gamma)}$  be instances of the void complex  $\emptyset$ .*

*If  $\sigma \neq X$  but  $\overline{X} = \emptyset$ , we let  $\Psi_{Q(\sigma, \gamma)}$  and  $\Phi_{Q(\sigma, \gamma)}$  be instances of the empty complex  $\{\emptyset\}$ .*

*We say that  $Q(\sigma, \gamma)$  models the link of  $\sigma$  restricted to  $\gamma$ .*

**Comment:** Although the previous definition looks similar to that for  $\text{Lk}(\Psi_R, \sigma)$  on page 27, there are some differences: (a) Here, we require that  $\sigma$  be a simplex in  $\Psi_R$ . (b) Here, we do *not* assume  $\gamma = \phi_R(\sigma)$ , merely  $\gamma \subseteq \phi_R(\sigma)$ . (c) Finally, Definition 8 on page 27 creates an empty complex whereas the current definition creates a void complex when  $\sigma = X \in \Psi_R$ . In summary: When  $\sigma \neq X$ ,  $Q(\sigma, \gamma)$  models those simplices of  $\text{Lk}(\Psi_R, \sigma)$  that are witnessed by attributes in  $\gamma$ .

**Theorem 19** (Isotropy = Minimal Identification = Sphere). *Let  $R$  be a relation and suppose  $\emptyset \neq \gamma \in \Phi_R$ . Let  $\sigma = \psi_R(\gamma)$ . Then the following four conditions are equivalent:*

- (a)  $\gamma$  is isotropic.
- (b)  $\gamma$  is minimally identifying (for  $\sigma$ ).
- (c)  $\Psi_{Q(\sigma, \gamma)} \simeq \mathbb{S}^{k-2}$ , with  $k = |\gamma|$ .
- (d)  $\Phi_{Q(\sigma, \gamma)} = \partial(\gamma)$ .

See Appendix F.3 for a proof.

**Collaboration Example Revisited:** To illustrate Theorem 19, consider again the example of Figure 23. Recall that together the travel guides for BERLIN and DUBLIN identify **Claire**. Indeed,  $\{\text{BERLIN}, \text{DUBLIN}\}$  is a minimally identifying set of books for **Claire**. It is isotropic, as Figure 24 shows. Figure 29 depicts the link of **Claire** restricted to  $\{\text{BERLIN}, \text{DUBLIN}\}$ , modeled by relation  $Q'$ . Observe that  $\Phi_{Q'} = \partial(\{\text{BERLIN}, \text{DUBLIN}\})$  and that  $\Psi_{Q'} \simeq \mathbb{S}^0$ , as the theorem asserts.

## 10.6 Poset Lengths and Information Release

We have seen how minimal identification appears topologically via spheres. Spheres are isotropic so perhaps it is not surprising that they encode isotropic attribute release sequences. We cannot therefore expect a spherical characterization for the problem of finding a maximally long informative attribute release sequence. Instead, we find an answer in the combinatorial structure of the doubly-labeled poset  $P_R$  and its lattice  $P_R^+$ . We summarize the key results below. For proofs, see Appendix F.

**Lemma 20** (Informative Attributes from Maximal Chains). *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. Suppose  $\{(\sigma_k, \gamma_k) < \dots < (\sigma_1, \gamma_1) < (\sigma_0, \gamma_0)\}$ , with  $k \geq 1$ , is a maximal chain in  $P_R^+$ .*

*Define  $y_1, \dots, y_k$  by selecting some  $y_i \in \gamma_i \setminus \gamma_{i-1}$ , for each  $i = 1, \dots, k$ .*

*Then  $y_1, \dots, y_k$  is an informative attribute release sequence for  $R$ .*

*Moreover,  $(\phi_R \circ \psi_R)(\{y_1, \dots, y_i\}) = \gamma_i$  for each  $i = 0, 1, \dots, k$ .*

(Notes: (a) For a maximal chain,  $\gamma_k = Y$  and  $\sigma_0 = X$ . (b) The hypothesis  $k \geq 1$  excludes any relation  $R$  for which  $\hat{0}_R = \hat{1}_R$ .)

Lemma 20 implies that every maximal chain in the doubly-labeled poset associated with a relation gives rise to an informative attribute release sequence that tracks the chain.

A partial converse holds as well:

**Lemma 21** (Chains from Informative Attributes). *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. Suppose  $y_1, \dots, y_k$  is an informative attribute release sequence for  $R$ , with  $k \geq 1$ .*

*Let  $\gamma_i = (\phi_R \circ \psi_R)(\{y_1, \dots, y_i\})$  and  $\sigma_i = \psi_R(\gamma_i)$ , for  $i = 1, \dots, k$ .*

*Then  $\{(\sigma_k, \gamma_k) < \dots < (\sigma_1, \gamma_1) < (X, \gamma_0)\}$  is a (not necessarily maximal) chain in  $P_R^+$ , with  $\gamma_0 = \phi_R(X)$ .*

Consequently one can obtain all informative attribute release sequences as subsequences of those constructed from maximal chains in  $P_R^+$ .

**Comment about “length”:** The *length*  $\ell(P)$  of a poset  $P$  is defined to be one less than the number of elements in the longest chain of the poset [18]. The *length* of an informative attribute release sequence  $y_1, \dots, y_k$  is  $k$ . These definitions match much like the dimension of a simplex is one less than the number of its elements.

**Corollary 22** (Maximal Length). *The maximum length of an informative attribute release sequence for a relation  $R$  is  $\ell(P_R^+)$ .*



**Corollary 23** (Maximal Identification Length). *Suppose  $R$  is a relation such that no attribute is shared by all individuals. For any  $(\sigma, \gamma) \in P_R$ ,  $r_{\text{slow}}(\sigma) = \ell(P_{Q(\sigma, \gamma)}) + 2$ .*

**Collaboration Example Re-Revisited:** Returning again to the travel guide example, observe in Figure 25 that  $\ell(P_G^+) = 4$ . This tells us, by Corollary 22, that a longest informative attribute release sequence for relation  $G$  contains four attributes. Indeed, we can pick three attributes to identify an individual, and then a fourth to form an inconsistency. How do we know that we can choose three attributes informatively to identify an individual? For example,  $\text{Lk}(\Psi_R, \text{Claire})$  is shown in Figure 26 with associated lattice  $P_C^+$  in Figure 27. In this case  $\ell(P_C) + 2 = \ell(P_C^+) = 3$ . Moreover, by the construction of Lemma 20, we can read off four different such informative sequences, namely the first four sequences appearing in Figure 24.

We thus see that  $r_{\text{slow}}(\text{Claire}) = 3$ , and as we have seen previously,  $r_{\text{fast}}(\text{Claire}) = 2$ . In other words, if **Claire** has control over how to release information, she can draw out identification for three books, while the fastest anyone can identify her is via two books.

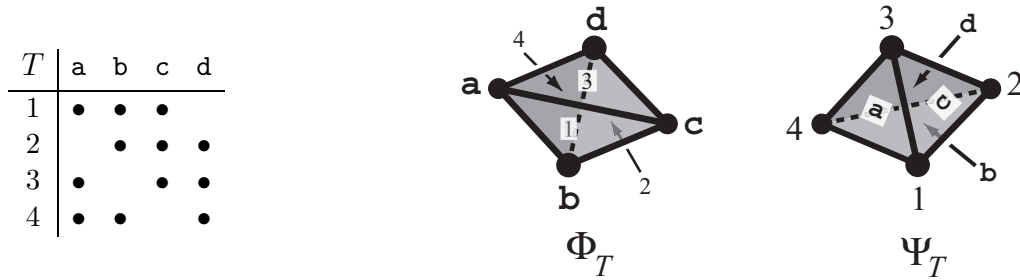


Figure 30: Relation  $T$  describes four individuals with four attributes, with Dowker complexes that are boundary complexes of tetrahedra, meaning they have homotopy type  $\mathbb{S}^2$ .

In contrast, consider the tetrahedral relation of Figure 30. The Dowker complexes are boundary complexes, so we know that no attribute or association inference is possible. This is evident from the lattice  $P_T^+$  depicted in Figure 31 as well. It has length 4, just as did the travel guide lattice, but the inference structure is now different. For any  $(\sigma, \gamma) \in P_T$ , with  $Q = Q(\sigma, \gamma)$  modeling  $\text{Lk}(\Psi_T, \sigma)$  on attributes  $\gamma$ , we see that  $\Phi_Q = \partial(\gamma)$  and thus that  $\ell(P_Q^+) = \ell(P_Q) + 2 = |\gamma|$ . This tells us, by Theorem 19 and Corollary 23, that  $r_{\text{fast}}(\sigma) = r_{\text{slow}}(\sigma) = |\gamma|$ , as one would expect in an inference-free world. For a specific instance, Figure 32 describes  $Q = Q(\{3\}, \{a, c, d\})$  along with  $Q$ 's Dowker complexes and the lattice  $P_Q^+$ .

## 10.7 Hidden Holes

We saw via Theorem 19 that whenever a set of attributes  $\gamma$  minimally identifies some set of individuals  $\sigma$ , then the link of  $\sigma$ , restricted to those simplices that are witnessed by attributes in  $\gamma$ , defines a sphere in both Dowker complexes. It is a hole.

All sets of individuals that are identifiable in some way, in other words, that appear in the doubly-labeled poset  $P_R$  of a relation, must be minimally identifiable in some way. That suggests there must be holes everywhere in a relation's Dowker complexes, and yet we do not see many holes. What is going on?

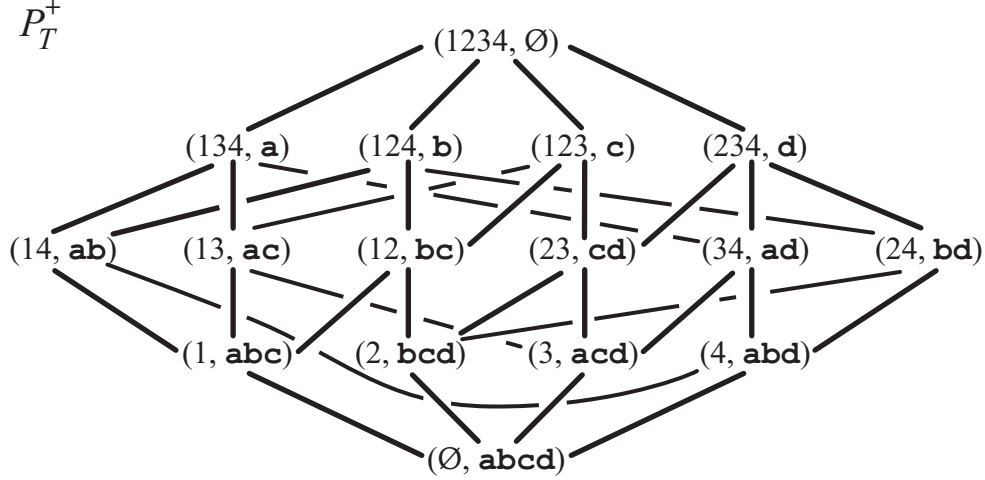


Figure 31: The lattice  $P_T^+$  for the tetrahedral relation of Figure 30. Each element is a pair of sets  $(\sigma, \gamma)$  such that  $\sigma = \psi_T(\gamma)$  and  $\gamma = \phi_T(\sigma)$ . (We have elided commas and braces in sets, for ease of viewing.) The lattice is isomorphic to the Boolean algebra on four elements, consistent with the fact that  $T$  preserves both association and attribute privacy. If one removes the top and bottom elements, the remaining poset  $P_T$  has  $\mathbb{S}^2$  homotopy type.

The answer is that the restricted link construction  $Q(\sigma, \gamma)$  focuses on a particular subrelation, thereby highlighting the hole. The hole itself could be hidden in the encompassing relation. For instance, we saw that relation  $Q$  of Figure 32 defines an  $\mathbb{S}^1$  hole. If  $Q$  happened to be a subrelation of relation  $R$  as in Figure 33, then  $Q$  would not be a hole when viewed in  $R$ , merely a boundary.

Notice that the lattice  $P_R^+$  is isomorphic to the lattice  $P_Q^+$ . The difference is that for every lattice element  $(\sigma, \gamma)$ , the set of individuals  $\sigma$  includes 3 in  $P_R^+$  but not in  $P_Q^+$ . Consequently, the bottom element  $(3, \text{acd})$  of  $P_R^+$  is actually an element of the poset  $P_R$ , meaning  $\Delta(P_R)$  is a cone, hence contractible. In contrast, the poset  $P_Q$  does not contain the bottom element  $(\emptyset, \text{acd})$  of  $P_Q^+$  and so  $\Delta(P_Q)$  has  $\mathbb{S}^1$  homotopy type.

**Aside:** Why not always focus on a relation's lattice rather than its doubly-labeled poset? Because the lattice is always contractible. Any interesting topology lies in the poset. See [18].

**Conclusion:** Even though  $R$  is contractible, it offers the same choices for informative attribute release sequences as does  $Q$ . More generally, the analysis of this subsection suggests that one look for holes in *subrelations* of a given relation. Looking at links is one way to focus on subrelations. Removing individuals or attributes that represent cone apexes is another, as we just saw. More generally, any simplicial cycle that can be represented by a subrelation defines a useful hole even though the hole appears to be filled-in. So long as one can remove any coboundary of that cycle, *by restricting the relation without destroying the cycle*, the cycle is informational. In particular, it offers opportunities for informative attribute release sequences, as the next subsection makes precise.

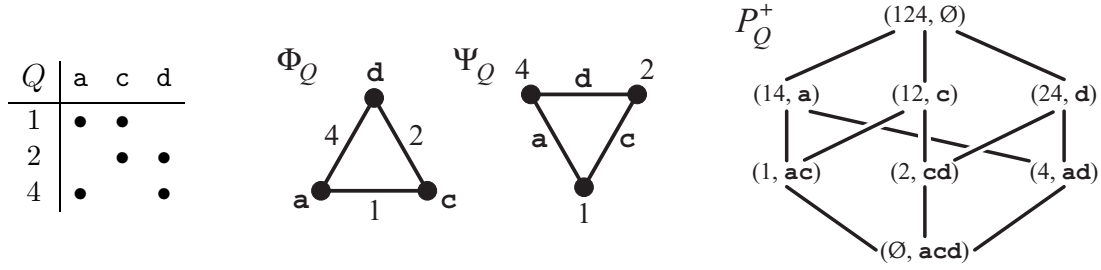
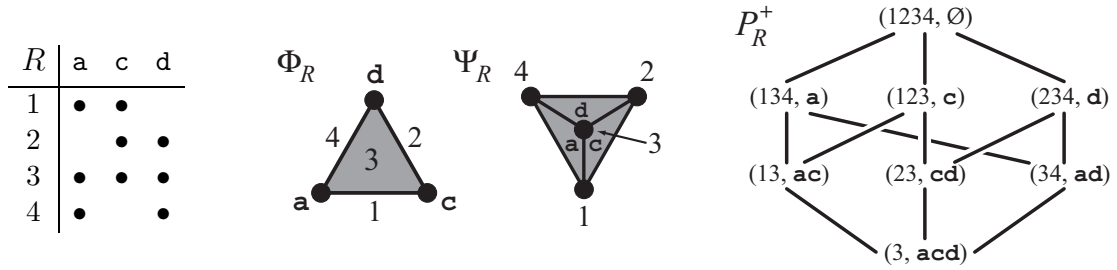
Figure 32: Relation  $Q$  models  $\text{Lk}(\Psi_T, 3)$ , with  $T$  as in Figure 30.

Figure 33: Relation  $R$  fills in the hole of relation  $Q$  from Figure 32. It is still true that  $Q$  models a link, namely  $\text{Lk}(\Psi_R, 3)$ .  $R$  and  $Q$  have the same lattice structure, but the bottom element of  $P_R^+$  defines the set of individuals  $\{3\}$ , whereas the bottom element of  $P_Q^+$  defines the empty set. Thus relation  $R$  defines a contractible poset for  $P_R$ , whereas relation  $Q$  defines an  $\mathbb{S}^1$  hole for  $P_Q$ .

## 10.8 Bubbles are Lower Bounds for Privacy

We have seen minimal identifiability characterized by holes, via Theorem 19. The previous subsections make clear that the topological characterization of  $r_{\text{slow}}$  is not so direct. In this subsection we establish a sufficient condition. We will see that holes provide lower bounds for  $r_{\text{slow}}$ . We will focus on a relation and its links, but these results apply more generally to any hidden holes made visible by focusing on subrelations, as outlined in the previous subsection.

The connection between a relation's poset  $P_R$  and its lattice  $P_R^+$  suggests the following:

**Definition 24** (Almost a Join-Based Lattice). *Let  $P$  be a finite poset. We say that  $P$  is almost a join-based lattice if adjoining a top element  $\hat{1}$  means  $P \cup \{\hat{1}\}$  is a join semi-lattice.*

**Comments:** (a) We adjoin  $\hat{1}$  even if  $P$  already has a top element. (b) Since  $P$  is finite, if  $P$  is almost a join-based lattice, then if we adjoin both a top element  $\hat{1}$  and a bottom element  $\hat{0}$ , the result will be a lattice. See also [18].

This definition leads our key insight (for a proof, see Appendix G):

**Theorem 25** (Many Chains). *Let  $P$  be almost a join-based lattice. Suppose  $P$  has reduced integral homology in dimension  $k \geq 0$ , that is,  $\tilde{H}_k(\Delta(P); \mathbb{Z}) \neq 0$ .*

*Then there are at least  $(k+2)!$  maximal chains in  $P$  of length at least  $k$ .*

**Interpretation:** The theorem says that a homology hole acts at least as powerfully as a spherical hole, from the perspective of producing informative attribute release sequences. Consider again the tetrahedral relation of Figure 30. The Dowker complexes form two-dimensional holes, so  $k = 2$  and  $(k + 2)! = 24$ . The poset  $P_T$  is the proper part of the lattice shown in Figure 31, that is, all the elements except the topmost and bottom-most. There are indeed 24 different chains of length 2, i.e., containing three elements, in  $P_T$ .

These chains represent the 24 different ways in which one might start at a vertex of one of the Dowker complexes, walk from that vertex to the middle of an incident edge, then walk from the middle of that edge to the centroid of an encompassing triangle. For instance: the walk from the vertex  $\{a\}$  to the edge  $\{a, c\}$  to the triangle  $\{a, c, d\}$  in  $\Phi_T$ . One can think of this walk as sequential acquisition of attribute information about an individual in a particular order. The order may perhaps be determined by chance or perhaps by an individual purposefully releasing information in a particular order. Once (and only once) one has arrived at the center of the triangle, has one fully identified the individual (in this case, as individual #3).

With that observation, we finally see how the global geometry / topology of the Dowker complexes, as encoded in their common poset, affects inference, beyond the local simplicial collapses of the closure operators. We will presently formalize this insight via two corollaries to Theorem 25.

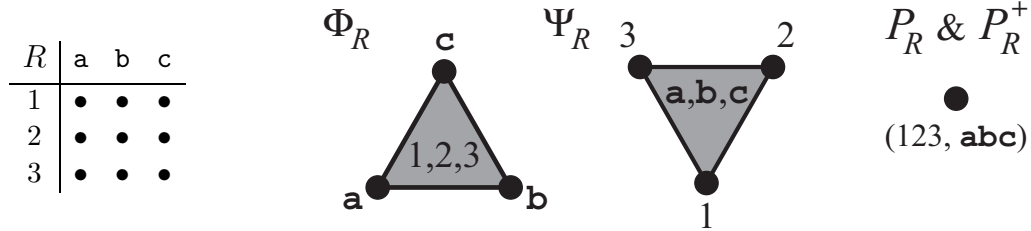


Figure 34: Relation  $R$  describes three individuals all of whom have the exact same three attributes. The Dowker complexes are both triangles, but the poset  $P_R$  is a single point. This single point captures the indistinguishability of the individuals and the attributes. In fact,  $P_R^+ = P_R$ , meaning one can infer everything from nothing.

We caution that the dimension of a simplex in a Dowker complex is not meaningful in and of itself, since the simplex may collapse under the closure operators. (Consider the example of Figure 34, in which the Dowker complexes are full triangles, but the doubly-labeled poset is a single point.) Instead, the length of chains in a relation's poset is significant. Holes prevent these chains from being short.

**Corollary 26** (Holes Reduce Inference). *Let  $R$  be a relation. Suppose  $P_R$  has reduced integral homology in dimension  $k \geq 0$ . Then there are at least  $(k + 2)!$  maximal chains in  $P_R$  of length at least  $k$ .*

**Corollary 27** (Holes Defer Recognition). *Let  $R$  be a relation and let  $(\sigma, \gamma) \in P_R$ .*

*Define  $Q = Q(\sigma, \gamma)$  as per Definition 18 and recall Definition 16, from pages 42–43.*

*Suppose  $P_Q$  has reduced integral homology in dimension  $k \geq 0$ .*

*Then there are at least  $(k + 2)!$  distinct informative attribute release sequences  $y_1, \dots, y_\ell$  for  $R$ , each with  $\ell \geq k + 2$ , such that  $\psi_R(\{y_1, \dots, y_\ell\}) = \sigma$ . Consequently,  $r_{\text{slow}}(\sigma) \geq k + 2$ .*

**Comment:** Since  $(\sigma, \gamma) \in P_R$  and by the homology hypothesis, relation  $Q(\sigma, \gamma)$  models link  $\text{Lk}(\Psi_R, \sigma)$ .

**Collaboration Example Once Again:** The Dowker complexes for the travel guide example of Figure 23 have  $\mathbb{S}^1$  homotopy type, meaning  $P_G$  has homology in dimension  $k = 1$ . Corollary 26 therefore says that there are at least 6 maximal informative attribute release sequences in  $P_G$ . Being maximal, each must identify some author. In fact, we saw that there were 4 different maximal informative attribute release sequences for identifying any one author. Since there are 5 authors,  $P_G$  actually contains at least 20 distinct maximal informative attribute release sequences. Can we find more via our corollaries? Not directly for individual authors, since, as we saw via Figure 26, the link of any one author is contractible, meaning that Corollary 27 does not help us directly.

There is more to be said however: The proof of Theorem 25 actually establishes that, for certain representatives of a homology class, the maximal elements in the support of that representative give rise to  $(k + 1)!$  chains. In the collaboration example, by choosing the homology generator appropriately, this implies that for each author there are at least two informative attribute release sequences for identifying the author. That gives us 10 sequences overall for relation  $G$ . To find 20, we would have to examine links of pairs of authors. There are 10 such links, 5 of which look similar to the one in Figure 28 on page 41. Each is an instance of  $\mathbb{S}^0$ , meaning each has two different identifying iars. That therefore gives us 10 iars for identifying pairs of authors, and thus 20 iars for identifying individual authors. — Corollary 27 further allows us to conclude that the maximal length of an informative attribute release sequence that identifies a given pair of authors is at least two. Consequently, the maximal length of an informative attribute release sequence that identifies a given individual author must be at least (and thus exactly) three. — Observe that one can draw these various conclusions guided by homology.

## 11 Experiments

An individual may wish to reveal information about himself/herself while delaying full identification. We saw in Section 10.8 that homology provides a lower bound on the number and length of such informative attribute release sequences. The lower bound need not be tight. In order to test the existence of the lower bound as well as see that it is not tight, we examined two datasets of different character:

**Medals:** We obtained this dataset in August 2014 from

<http://www.tableausoftware.com/public/community/sample-data-sets>.

The dataset contains information about athletes who participated in the Olympics during the years 2000–2012. The attributes we considered were:

Age, Country, Year, Sport, Gold Medals, Silver Medals, Bronze Medals

(The last three attributes count the number of medals won by an athlete.)

Every athlete therefore has exactly 7 attributes, with each attribute taking on one of a finite discrete set of mutually exclusive values. We represented these 7 dimensions of multivalent attributes as a collection of 233 binary attributes.

There are 8613 individuals (we regarded the same athlete in different years as distinct individuals), who partition into 6955 equivalence classes (for team sports, athletes are often indistinguishable).

The result is a binary relation  $M$  with 6955 rows and 223 columns.

**Jazz:** We assembled this relation in June 2015 by examining the website

<http://www.redhotjazz.com>.

The website contains information about jazz musicians and bands, mainly from the early to late-mid 20th century.

We assembled a relation  $J$  whose rows are indexed by musicians and whose columns are indexed by bands, with  $(m, b) \in J$  meaning that musician  $m$  played in band  $b$ .

The result is a binary relation  $J$  with 4896 rows and 990 columns.

Cautions: We were not particularly careful to determine whether different spellings of a name really meant the same person. For some bands, the website listed one or more bandmembers as “unknown”. We ignored those bandmembers. We ignored bands for whom we could not determine any bandmembers. Since our goal was to understand how homology influences the existence of informative attribute release sequences, such noise in constructing a relation should not be particularly significant. If one wished to draw sociological conclusions about the spread of music, one would need to be more careful.

We encountered the jazz website because it was the source of data for a paper on collaboration networks [11] that considered the dual nature of individuals and attributes. That paper constructed two graphs, one with musicians as vertices and bands as edges, the other with those roles reversed. One can view those graphs as the 1-skeleta of our Dowker complexes.

The paper observed that drawing conclusions in one space might be easier than in the other, depending on the question being asked. The paper drew some conclusions about musical influence.

## 11.1 Compare and Contrast

We review some key differences between the two relations  $M$  and  $J$ .

**Identifiability:** The original 8613 individuals in the Olympic Medals dataset were not all uniquely identifiable. For some athletes, even knowing an athlete's full set of 7 attributes left ambiguity as to the athlete's identity. This was true for 2810 of the athletes. Fortunately, an athlete's ambiguity was fully symmetric, meaning that one could in fact partition the set of all athletes into equivalence classes. This symmetry was likely due to the fact that some competitions involved teams, with team members indistinguishable from each other. Each equivalence class then formed a uniquely identifiable "individual" in relation  $M$ .

For the Jazz relation, 863 of the 4896 musicians were uniquely identifiable, but 4033 were not. Unfortunately, this time the ambiguity was not fully symmetric. One could again partition the 4033 individuals into 1022 equivalence classes based on having identical rows in  $J$ . However, some rows remained subsets of other rows, giving a directionality to the ambiguity. For this reason, we did not pass to equivalence classes.

**Attribute Size:** In the medals relation  $M$ , every individual has exactly 7 attributes, describing one value for each of the 7 possible fields: **Age**, **Country**, **Year**, **Sport**, **Gold Medals**, **Silver Medals**, **Bronze Medals**. Consequently there are also always exactly 7 attributes in each link relation.

In the Jazz dataset, there was no structural bound to the number of bands in which a musician might have played, so a musician's attributes could be many. The largest number of bands in which any one musician played was in fact 44. The average was a little over 2 and the median 1. Conversely, the largest band had 288 musicians, with an average of 10.4 and a median of 7.

**Link Size:** For  $M$ , the number of other athletes in any given athlete's link was always close to the entire set of possible athletes. With only 7 attribute fields, any two athletes shared almost certainly some attribute value.

In contrast, for the 767 musicians in  $J$  for whom we computed links (described further in Section 11.4), the number of other musicians in any given musician's link was relatively low. The average was 55.3, the median 37, with a maximum of 301. With musicians generally playing in few bands, each encountered on average only a few score fellow musicians of the 4895 other musicians he/she might have encountered.

## 11.2 Homology Computations

For each of the link relations discussed below, we computed homology of the Dowker complex  $\Phi_Q$ , with relation  $Q$  modeling the link.<sup>3</sup> Since our goal was to find lower bounds for informative attribute release sequences, we modified  $\Phi_Q$  slightly, as suggested by Section 10.7. Specifically, whenever  $\Phi_Q$  was a cone with more than one maximal simplex, we removed all its cone apexes.

Note: The homology lower bound results of Section 10 and Appendix G do not depend directly on the chain coefficients being integers. We therefore computed homology with  $\mathbb{Z}_2$  coefficients, using the **Perseus** software previously written at the University of Pennsylvania: <http://www.sas.upenn.edu/~vnanda/perseus/>.

## 11.3 Homology and Release Sequences in the Olympic Dataset

**Overall Homology:** A collection of  $k$  multivalent discrete attributes produces Dowker complexes with homotopy types that are wedges of  $\mathbb{S}^{k-1}$ s, assuming that all possible combinations of the attributes are represented by individuals.

Consequently, with every individual having exactly 7 attributes, one might expect to see some homology in dimension 6. But of course, not every combination is possible. For instance, no one athlete is going to simultaneously win the gold, silver, and bronze medals in the same event. *From this perspective, real-world constraints show up as absence of potential homology.* In fact, relation  $M$  has the Betti numbers described in Table 2, computed using  $\mathbb{Z}_2$  coefficients.

$d$	0	1	2	3	4
$\beta_d$	1	0	23	757	503

Table 2: Betti numbers for the topology of the Olympic Medals relation.

The table suggests that there are quite a few informative attribute release sequences of length at least 5 for identifying athletes.

**Link Homology:** We computed the link of each athlete in  $M$ , and determined homology for the resulting relation, with the proviso mentioned above. Specifically, we removed all cone apexes from an athlete’s Dowker complex  $\Phi_Q$  (assuming it contained more than one maximal simplex) before computing homology, with  $Q$  being the link relation. Of the 6955 links, 3822 contained attribute cone apexes in  $\Phi_Q$ .

Table 3 summarizes the results. One may conclude more strongly now that (at least) 2198 athletes each have (at least) 120 different ways of releasing (at least) 5 of their 7 attributes in ways that do not fully identify the athlete before those 5 attributes have been released.

**Informative Attribute Release Sequences:** We computed a maximal length informative attribute release sequence for each link relation. One can find such a sequence by searching for a least-cost path from  $\hat{1}_Q$  to  $\hat{0}_Q$  in  $P_Q^+$ , picking attributes along the way as per the construction of Lemma 21 on page 44, with cost being the number of attributes inferred as one traverses the path. Here  $Q$  is again the link relation. Of the 6955 athletes, 6229 actually had a maximal informative attribute release sequence of length 7. Each such athlete could order his/her

<sup>3</sup>Formally, the link is equal to  $\Psi_Q$ . By Dowker’s Theorem,  $\Psi_Q$  and  $\Phi_Q$  have the same homology.



$d$	0	1	2	3	4
# of athletes	229	1355	2773	2198	57
$\max_{\text{athletes}} \beta_d$	2	4	7	4	2

Table 3: Histogram indexed by dimension  $d$ , describing athletes whose links  $\text{Lk}(\Psi_M, \text{athlete})$  have homology in dimension  $d$  (after removal of attribute cone apexes from the dual complexes), for the 6955 athletes in the Olympic Medals relation  $M$ . Also shown are the maximum Betti numbers seen in each dimension, with the maximum taken over all possible athletes.

attributes in such a way that his/her identity does not become fully known until s/he has released all 7 attributes. Of the remaining athletes, 719 had a maximal informative attribute release sequence of length 6, and 7 had a maximal length of 5.

Of course, Corollary 27 makes a stronger claim, suggesting permutability of attributes. Consequently, we computed for each link relation all possible isotropic sets of attributes (see again Definition 14 on page 41, now with  $Q$  in place of  $R$ ). Table 4 summarizes the results.

$ \kappa $	2	3	4	5	6
# of athletes	6955	6955	6955	5568	171
$\max_{\text{athletes}}  \{\kappa\} $	21	35	35	21	5

Table 4: Histogram indexed by size  $|\kappa|$ , describing athletes whose link relations contain isotropic attribute sets  $\kappa$ . An athlete may have several distinct (possibly overlapping) such sets for any given size. Also shown therefore are the maximum numbers of such sets, with the maximum taken over all possible athletes. For example: 171 athletes have at least one isotropic set of size 6 in their link relation, and the maximum number of such sets any one athlete has is 5.

**Scatterplot:** Finally, we computed for each link a pair of numbers  $(h, i)$ , with  $h$  representing a measure of link homology and  $i$  representing a measure of informative attribute release sequences for the link relation. The resulting scatterplot appears in Figure 35. One can see that homology acts as a lower bound for informative attribute release sequences.

The exact formulas for  $h$  and  $i$  are not that significant, but we mention them here for completeness. To get a measure of homology, we assembled for each link a vector with the Betti numbers computed earlier:  $(\beta_0, \beta_1, \beta_2, \beta_3, \beta_4)$ . We looked at all such vectors to determine maximum values for each component (as given in Table 3). We could then think of the vector as defining, in reverse order, a varying-radix numeral. We converted that numeral to an integer. So, for instance, if a given link were to have Betti vector  $(1, 3, 5, 0, 0)$ , then its  $h$  value would be  $1 + 3 \cdot (2 + 1) + 5 \cdot (4 + 1) \cdot (2 + 1) = 85$ . Observe that the  $h$  value for a contractible link is 1. In order to graph the scatterplot nicely, we scaled the  $h$ -axis by taking a fourth root.

We computed a link's  $i$  value similarly, now from the following vector of data:  $(\ell_{\max}, c_2, c_3, c_4, c_5, c_6)$ . Here  $\ell_{\max}$  is the largest  $\ell$  in an informative attribute release sequence  $y_1, \dots, y_\ell$  for the link relation, while  $c_k$  is the number of different isotropic attribute sets  $\kappa$  in the link relation such that  $|\kappa| = k$ . We scaled the  $i$ -axis by taking a logarithm.

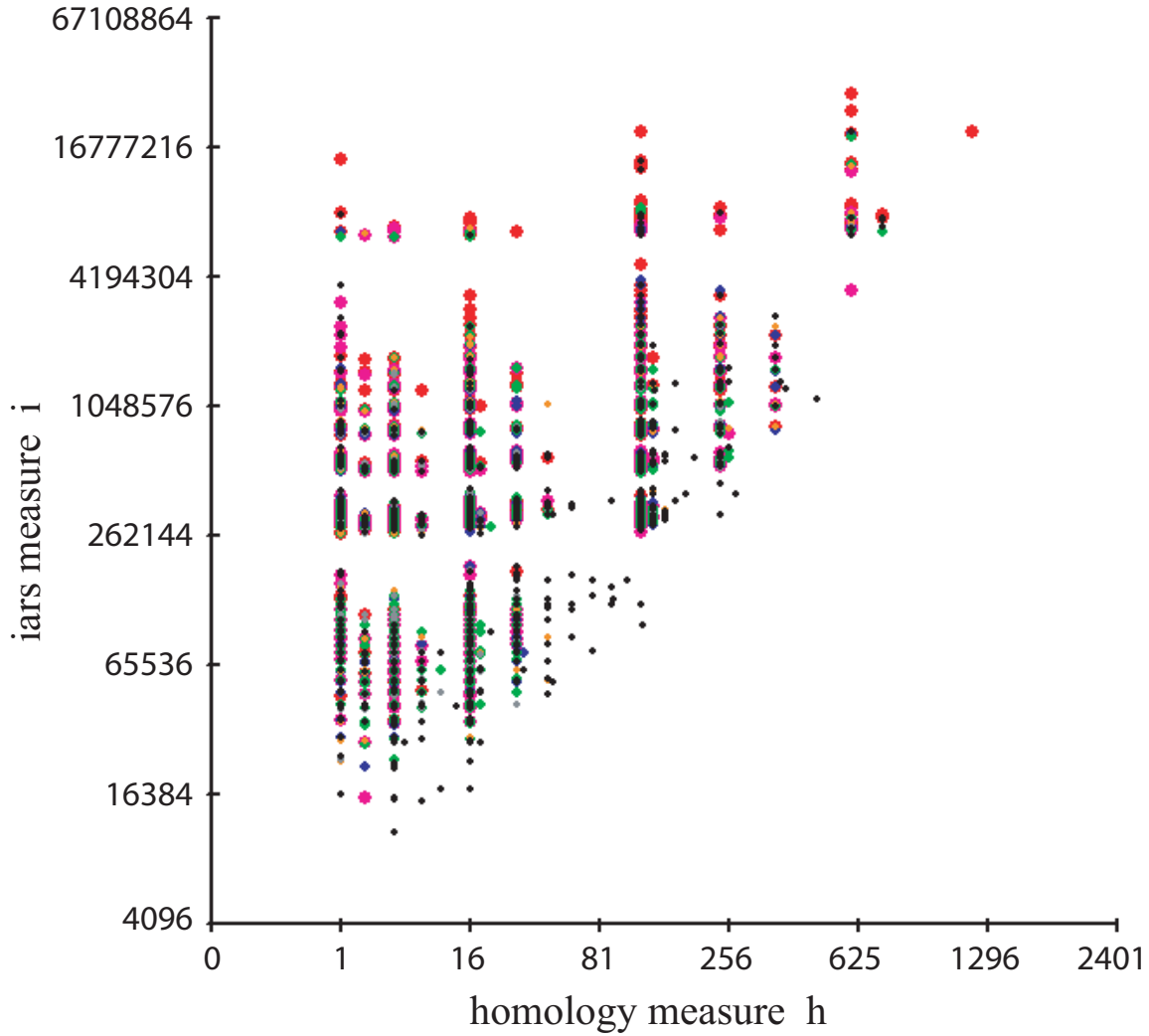


Figure 35: Scatterplot describing each athlete's link in the medals relation  $M$ . The scatterplot shows for each link a point  $(h, i)$ , with  $h$  a measure of the link's homology (after removal of attribute cone apexes) and  $i$  a measure of how many significant informative attribute release sequences exist for the link relation. The scatterplot underscores how homology is a lower bound for informative attribute release sequences, as described in Corollary 27.

(The colors and radii indicate the numbers of athletes in the links. The color ordering and size boundaries are:

BLACK—6821—SILVER—6831—ORANGE—6851—GREEN—6859—BLUE—6865—MAGENTA—6872—RED.

In this figure, the boundaries between colors were chosen so that each bucket holds roughly 1000 links. As one can see, the number of athletes in a link is generally large.)

### 11.4 Homology and Release Sequences in the Jazz Dataset

**Overall Homology:** Given the large number of bands in which some musicians played, and given memory constraints of our laptop at the time, we were not able to compute homology for the full Jazz relation  $J$ . Instead, we were able to compute the homology for restricted relations consisting of musicians that played in fewer than 20 bands. This covered 4856 of the 4896 musicians in the overall relation. Since we did not see any homology above dimension 2 in several of these restricted cases, we considered the 3-skeleton of  $\Phi_J$  as proxy for the topology of the full relation  $J$  and computed its homology. Table 5 summarizes the results. We verified using a graph algorithm that the full relation  $J$  did indeed have 107 components, as indicated by  $\beta_0$  for the 3-skeleton  $\Phi_J^{(3)}$ . Given the low amount of homology and the low dimension of such homology,  $J$  might not be telling us much about the length of informative attribute release sequences for the various musicians, suggesting we look at links.

$b$	$\Sigma$	$m$	$\beta_0$	$\beta_1$	$\beta_2$	$\beta_3$
14	$\Phi_{J b}$	4819	111	613	20	0
15	$\Phi_{J b}$	4831	111	613	32	0
16	$\Phi_{J b}$	4838	111	605	42	0
17	$\Phi_{J b}$	4848	110	603	58	0
18	$\Phi_{J b}$	4851	110	603	65	0
19	$\Phi_{J b}$	4856	109	596	75	0
$\infty$	$\Phi_J^{(3)}$	4896	107	550	93	—
15	$\Phi_{J'}$	767	18	595	32	0

Table 5: Betti numbers for subcomplexes  $\Sigma$  of  $\Phi_J$ , with  $J$  being the Jazz relation. The first six rows correspond to restrictions of  $J$  to musicians who played in at most  $b$  bands. For each row,  $m$  indicates the number of musicians in the relation. The penultimate row describes the 3-skeleton of  $\Phi_J$ . The last row refers to a relation  $J'$  described further in the text.

**Link Homology:** We computed the link of some of the musicians in  $J$ , and determined homology for the resulting relations (again after removal of attribute cone apexes). Table 6 summarizes the results. Given the inability to fully identify some musicians even knowing all their bands (as described in Section 11.1) and the difficulty of computing homology for large band memberships, we computed links only for a subset of the musicians. We required each musician to be uniquely identifiable, to have played in at most 15 bands, and to have a nontrivial link. There were 767 such musicians. Betti numbers for the relation  $J'$  representing the restriction of  $J$  to these 767 musicians also appear in Table 5. (Note, however, that we computed the full link  $\text{Lk}(\Psi_J, \text{musician})$  for each of the 767 musicians, not merely  $\text{Lk}(\Psi_{J'}, \text{musician})$ .) We removed attribute cone apexes from the link relation for 106 of these 767 musicians.

These results suggest that the relationships to other musicians do indeed *not* have many holes in them. Recall, from a topological perspective, one can assert the existence of at least  $(k+2)!$  distinct informative attribute release sequences of length at least  $k+2$  for any musician with a  $k$ -dimensional hole. For almost all musicians this means 2 sequences of length 2, for

$d$	0	1	2	3
# of musicians	604	145	20	1
$\max_{\text{musicians}} \beta_d$	7	6	3	1

Table 6: Histogram indexed by dimension  $d$ , describing musicians whose links  $\text{Lk}(\Psi_J, \text{musician})$  have homology in dimension  $d$  (after removal of attribute cone apexes from the dual complexes), for the 767 musicians who are uniquely identifiable in  $J$ , played in at most 15 bands, and had nontrivial link. Also shown are the maximum Betti numbers seen in each dimension, with the maximum taken over the 767 possible musicians. For  $d = 0$ , this means that 604 of the 767 musicians have connections to other musicians that split into disjoint groups. The maximum number of such disjoint components for any one musician is 7.

some it means 6 sequences of length 3, for a few it means 24 sequences of length 4, and for one musician it means 120 sequences of length 5. These observations are roughly in line with the actual data for informative attribute release sequences described next, though, as expected for the theoretical reasons discussed earlier, they constitute lower bounds.

**Informative Attribute Release Sequences:** We computed a maximal length informative attribute release sequence for each link relation. Table 7 summarizes the results. We mention in passing: Any attribute release sequence that is informative for a link relation is also informative for the encompassing relation. For a few musicians, the maximal sequence found within the link relation  $Q$  could be further extended in the encompassing relation  $J$ , with a prefix of one attribute, namely an attribute shared by all members of the link, yet remain informative and identifying. This occurred for the 17 musicians whose maximum sequence length  $\ell$  was 1.

We also computed for each link relation all possible isotropic sets of attributes. Table 8 summarizes those results.

$\ell$	1	2	3	4	5	6	7	8	9	10	11
# of musicians	17	248	218	125	72	35	23	15	11	2	1

Table 7: Histogram of musicians, indexed by length  $\ell$  of the longest informative attribute release sequence for the musician’s link relation, for the 767 musicians described in the text.

$ \kappa $	2	3	4	5
# of musicians	750	219	49	3
$\max_{\text{musicians}}  \{\kappa\} $	105	202	40	2

Table 8: Histogram indexed by size  $|\kappa|$ , describing musicians whose link relations contain isotropic attribute sets  $\kappa$ . Also shown are the maximum numbers of such sets, with the maximum taken over the 767 possible musicians described in the text.

**Scatterplot:** We computed for each link a pair of numbers  $(h, i)$ , with  $h$  representing a measure of homology and  $i$  representing a measure of the link’s informative attribute release sequences, much as for the medals relation  $M$  of Section 11.3. The resulting scatterplot appears in Figure 36. Again homology acts as a lower bound for informative attribute release sequences.

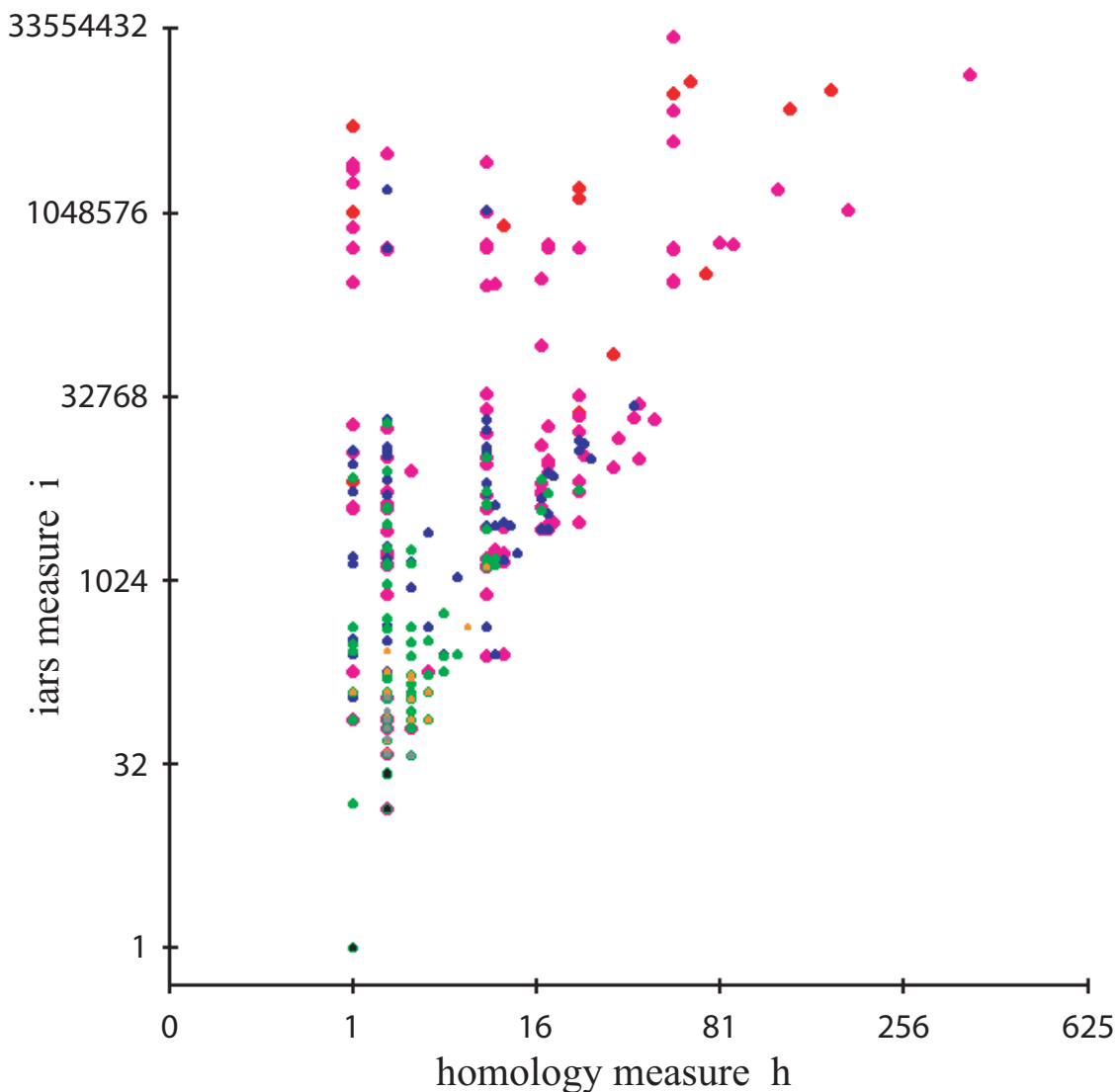


Figure 36: Scatterplot describing the links computed for 767 of the musicians in the Jazz relation  $J$ . The scatterplot shows for each link a point  $(h, i)$ , with  $h$  a measure of the link's homology (after removal of attribute cone apexes) and  $i$  a measure of the link's informative attribute release sequences.

(The colors and radii indicate the numbers of musicians in the links. Link sizes are fairly small. The color ordering and size boundaries are:

BLACK-5-SILVER-10-ORANGE-20-GREEN-50-BLUE-100-MAGENTA-200-RED.

In this figure, the buckets may hold noticeably varying numbers of links.)

## 12 Inference in Sequence Lattices

We have seen how a relation gives rise to a lattice via the Galois connection, as per Definition 12 on page 38. The lattice structure describes the ways in which privacy may be preserved or lost. Consequently, when thinking about privacy, one may be able to start with a lattice that does not necessarily arise directly from a relation.

This section will look at inferences from sequences of observations. The next section examines strategy obfuscation in planning with uncertainty.

We should also recall some equivalences. Lattices are particular kinds of partially ordered sets (posets). Posets and simplicial complexes are topologically identical [18]. One can move back and forth between these representations while preserving homeomorphism type (see Appendix A). Furthermore, one may describe a simplicial complex by a relation in several different ways that preserve homotopy type, including ways in which one of the two resulting Dowker complexes is identical to the original simplicial complex. In short, one has three different categories of structures with which to think about privacy: relations, simplicial complexes, and lattices. One may start with any one representation and build the other two from that.

### 12.1 Sequence Lattices for Dynamic Attribute Observations

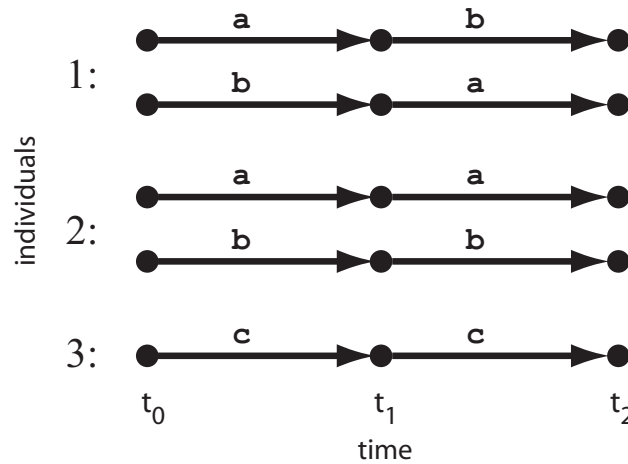


Figure 37: Three types of individuals and the attributes each might reveal in two successive time intervals.

Consider the dynamic process of Figure 37. The process models observations of individuals who reveal attributes over successive time steps. There are three possible individuals (or more generally, types of individuals). The first individual emits attributes “a” and “b” alternately at successive times, but one does not know which of those attributes one might see first. The second individual always emits the same attribute, either “a” or “b”, but one does not know *a priori* which it is. The third individual always emits the same attribute “c”.

A relation for these (types of) individuals that models the individuals in terms of single attributes appears as relation  $S$  in Figure 38. Individual #3 is distinguishable from the other

$S$	a	b	c
1	•	•	
2	•	•	
3			•

$T$	aa	bb	ab	ba	cc
1			•	•	
2	•	•			
3					•

Figure 38: Relation  $S$  describes individuals and single attributes, while  $T$  describes individuals and sequences of two attributes.

two individuals, but the relation provides no means for distinguishing those two individuals from each other. The relation is homogeneous with regard to single attributes for individuals #1 and #2. Of course, we can see from the dynamic process of Figure 37, that distinguishing information appears via sequences of two attributes. Relation  $T$  of Figure 38 models such sequences. Now all three individuals are uniquely identifiable. Should one wish to model inferences based on both one and two observations, one could use the relation  $S \cup T$ .

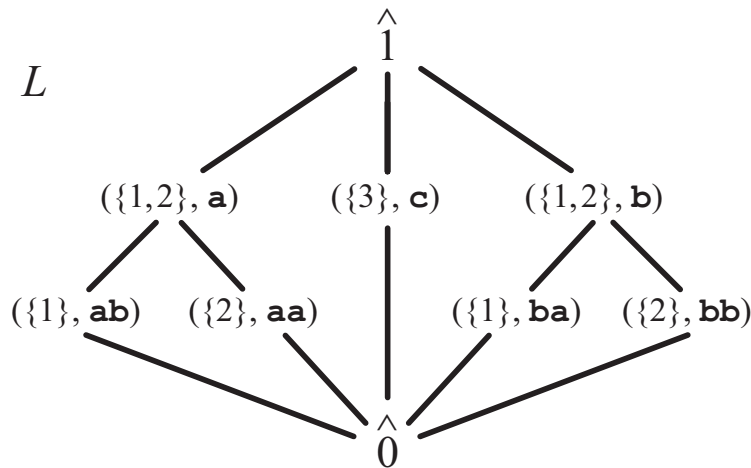


Figure 39: Lattice representing the dynamic process of Figure 37.

That jump from single to double attributes is useful, but where does it come from intrinsically? After all, without additional knowledge, we might simply consider infinitely long sequences, even though those would not add anything in this example. In fact, the dynamic process of Figure 37 gives us the information. It is itself basically a decision tree that amounts to the lattice of Figure 39. In that figure, we have annotated each internal node of the lattice with a pair, consisting of a set of individuals and either a single attribute or a sequence of two attributes. This lattice differs from previous ones in this report in that a set of individuals (or attributes more generally) is no longer constrained to appear in at most one node of the lattice. By allowing multiple nodes, we enhance our ability to encode state in the lattice. For example, observing attribute “a” carries different meaning depending on whether one has already seen attribute “a” or attribute “b” or no attribute at all. Also: While we could have included  $(\{3\}, cc)$  in the lattice, we did not need that element.

In the lattice of Figure 39 it is tempting to merge the two identifying nodes for individual

#1 into one node and to merge the two identifying nodes for individual #2 into one node. There is apparently no harm in doing so, in that the decision process would still be correct. However, the resulting structure would no longer be a lattice but merely a poset. That may or may not be desirable in a given application. For instance, using homology to estimate how long one can delay identification requires almost a join-based lattice to fulfill the hypotheses of Theorem 25 on page 47.

If we did want to merge nodes as just described, while maintaining a lattice, then we would perhaps also merge the two nodes containing the set  $\{1, 2\}$ , giving us the lattice of Figure 40. This lattice is similar to the lattice  $P_{S \cup V}^+$  that one would construct from the relation  $S \cup T$ , except that it does not include singleton attributes in the nodes identifying individuals #1 and #2 and it does not include the sequence “cc” in the node identifying individual #3.

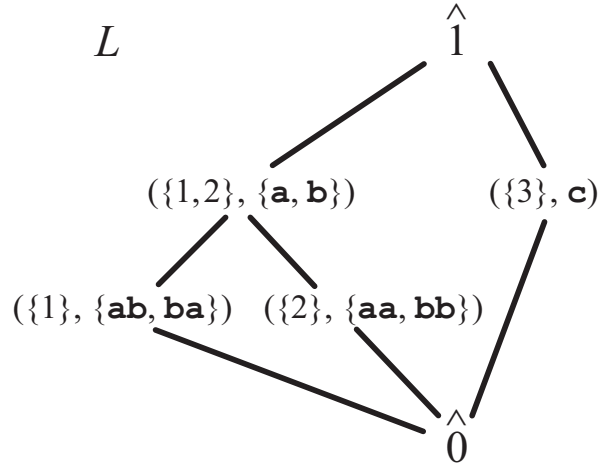


Figure 40: Modified lattice of Figure 39, after merging some nodes.

Regardless, the lattices of Figures 39 and 40 encode the inferences possible from the decision process of Figure 37. In particular, if we observe either attribute “a” or attribute “b”, then we know the set of possible individuals is  $\{1, 2\}$ ; we have excluded individual #3. Moreover, if we observe any two-attribute sequence, with attributes drawn from  $\{a, b\}$ , then we can identify the observed individual fully as either #1 or #2. Thus the required sequences come directly from the decision process, not requiring an intermediate representation as a relation.

## 12.2 Lattices of Stochastic Observations

The dynamic sequence perspective incorporates randomized response within the lattice framework. Instead of arising via a deterministic process as in Figure 37, the attributes “a” and “b” could flow from a stochastic process. One obtains an infinite lattice determined by increasingly longer sequences of observations. Depending on the confidence intervals one wishes to set, one obtains decision regions such as those sketched in Figure 41, with a central region of ambiguity, bounded by regions of exclusion, for identifying individuals.



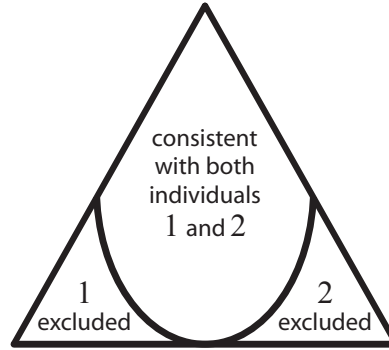


Figure 41: Sketch of an inference lattice for sequences of randomized response queries.

### 12.3 General Inference Lattices

Lattices are useful tools for inference. Rather than work with completely arbitrary lattices, we give here a definition that makes explicit the existence of two underlying structures over which we wish to perform inferences. However, we no longer assume a pair of underlying discrete spaces  $X$  and  $Y$  for individuals and attributes, but instead posit posets  $P$  and  $Q$ . The connection to our earlier relational perspective is that  $P$  would be the powerset of  $X$  and  $Q$  the powerset of  $Y$ . By allowing potentially different posets  $P$  and  $Q$  for a given lattice  $L$ , one can in some instances obtain different “views” of that lattice, thereby increasing flexibility in the interpretation process. For instance,  $Q$  might consist of all sequences up to a specified length or it might consist of *sets* of such sequences.

**Definition 28** (Inference Lattice). *Let  $P$  and  $Q$  be finite posets.*

*An inference lattice  $L$  with respect to  $P$  and  $Q$  is a bounded lattice whose proper part  $\bar{L}$  consists of pairs  $(p, q)$ , with  $p \in P$  and  $q \in Q$ , satisfying the following conditions:*

*For all  $(p_1, q_1)$  and  $(p_2, q_2)$  in  $\bar{L}$ :*

- (i)  $(p_1, q_1) \leq_L (p_2, q_2)$  if and only if  $p_1 \leq_P p_2$  and  $q_1 \geq_Q q_2$ ;*
- (ii)  $(p_1, q_1) \vee_L (p_2, q_2)$  is either  $\hat{1}_L$  or a pair  $(p, q) \in \bar{L}$  such that  $p$  is an upper bound for both  $p_1$  and  $p_2$  in  $P$  and  $q$  is a lower bound for both  $q_1$  and  $q_2$  in  $Q$ ;*
- (iii)  $(p_1, q_1) \wedge_L (p_2, q_2)$  is either  $\hat{0}_L$  or a pair  $(p, q) \in \bar{L}$  such that  $p$  is a lower bound for both  $p_1$  and  $p_2$  in  $P$  and  $q$  is an upper bound for both  $q_1$  and  $q_2$  in  $Q$ .*

*(Note that  $\hat{0}_L \leq_L (p, q) <_L \hat{1}_L$  for every  $(p, q) \in \bar{L}$ , given that  $\bar{L}$  is the proper part of  $L$ . Also be aware that  $\bar{L}$  need not, and generally will not, contain all possible pairs  $(p, q)$ .)*

**Inference Protocol:** Suppose we have observed some  $q \in Q$ . How should we interpret that observation in terms of the lattice  $L$ ? Here is a possible protocol: (In terms of our earlier relational model, one may view this protocol as inferring sets of individuals from sets of attributes.)

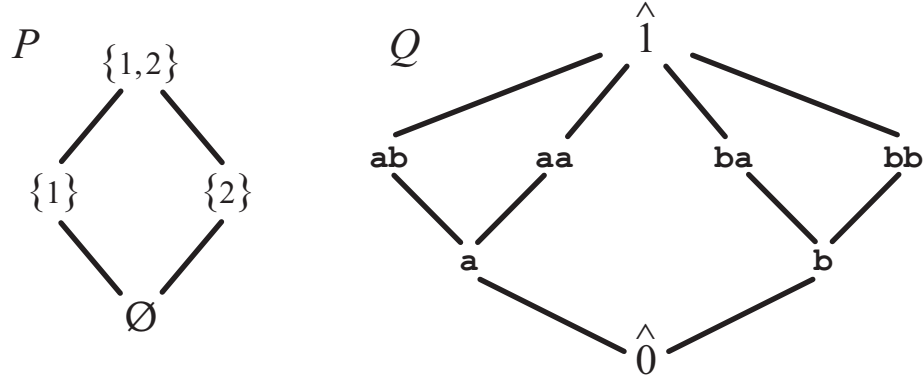


Figure 42: Poset  $P$  models some sets of individuals; poset  $Q$  models some sequences of attributes.

- Let  $\Gamma = \{(p', q') \in \bar{L} \mid q \leq_Q q'\}$ .
- If  $\Gamma = \emptyset$ , then we view  $q$  as inconsistent, implying interpretation  $\hat{0}_L \in L$ .
- Otherwise, let  $\Gamma_{\max}$  consist of all the maximal elements of  $\Gamma$  (maximal with respect to the partial order on  $L$ ). We view  $q$  as implying this set of elements in  $L$ . One can project each of those elements onto its  $P$  coordinate, if that is useful.

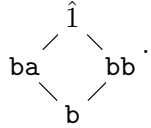
There is a dual protocol for interpreting an observation  $p \in P$ :  
(In terms of our earlier relational model, one may view this protocol as inferring sets of attributes from sets of individuals.)

- Let  $\Sigma = \{(p', q') \in \bar{L} \mid p \leq_P p'\}$ .
- If  $\Sigma = \emptyset$ , then we view  $p$  as inconsistent, implying interpretation  $\hat{1}_L \in L$ .
- Otherwise, let  $\Sigma_{\min}$  consist of all the minimal elements of  $\Sigma$  (minimal with respect to the partial order on  $L$ ). We view  $p$  as implying this set of elements in  $L$ . Again, one can project each of those elements onto its  $Q$  coordinate, if that is useful.

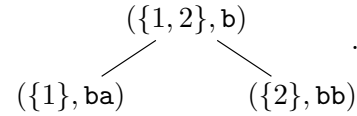
In our previous relational setting, the structure of Galois lattices ensured that each of  $\Gamma_{\max}$  and  $\Sigma_{\min}$  never contained more than one element. That need not be true for general inference lattices.

**Example:** Suppose  $P$  and  $Q$  are as in Figure 42. Here  $P$  models subsets drawn from the set of two individuals  $\{1, 2\}$ , while  $Q$  models sequential observations of “a” and “b”, of lengths one and two, as in our earlier example of Figure 37. The lattice  $L$  is as in Figure 39. For presentational simplicity, posets  $P$  and  $Q$  ignore individual #3 and attribute “c”, instead focusing on individuals  $\{1, 2\}$  and attributes  $\{a, b\}$ .

**Observing an attribute:** Suppose we have observed attribute “b”, i.e.,  $q = \mathbf{b}$ . What can we infer from  $q$  in  $P$  via  $L$ ? Let’s follow the protocol given above:

- The subposet of  $Q$  consisting of elements  $q'$  greater than or equal to  $q$  is: .

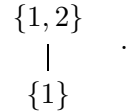
- Consequently,  $\Gamma$  is the following subposet of  $L$ :

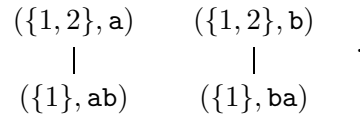


- There is one maximal element in  $\Gamma$ , so  $\Gamma_{\max} = \{(\{1, 2\}, \mathbf{b})\}$ .

Projecting onto the  $P$  component tells us how to interpret  $q$ : The observation “b” must have come from either individual #1 or individual #2. (This conclusion would hold as well if  $P$  had modeled individual #3 and if  $Q$  had modeled attribute “c”.)

**Observing an individual:** Suppose we have observed individual #1, i.e.,  $p = \{1\}$ . What can we infer from  $p$  in  $Q$  via  $L$ ? Again, let’s follow the inference protocol given earlier:

- The subposet of  $P$  consisting of elements  $p'$  greater than or equal to  $p$  is: .

- Consequently,  $\Sigma$  is the following subposet of  $L$ : .

- The minimal elements of  $\Sigma$  give us  $\Sigma_{\min} = \{(\{1\}, \mathbf{ab}), (\{1\}, \mathbf{ba})\}$ .

Projecting onto the  $Q$  component tells us how to interpret  $p$ : The individual observed can or did reveal one of the two-attribute sequences “ab” or “ba”.

**Comment:** The poset  $Q$  of Figure 42 would not be very useful for inferences in the lattice of Figure 40, since that lattice now models attribute observations as sets of sequences rather than merely as sequences. We would instead probably want  $Q$  to be something like the poset of Figure 43. So even though  $L$  has become simpler than in Figure 39,  $Q$  has become more complicated. On the other hand, the new  $(L, P, Q)$  triple means that one can infer  $(\{1, 2\}, \{\mathbf{a}, \mathbf{b}\})$  from the observation “b”. As before, that says the observation “b” must have come from individual #1 or #2, but it also says that the individual could alternatively have

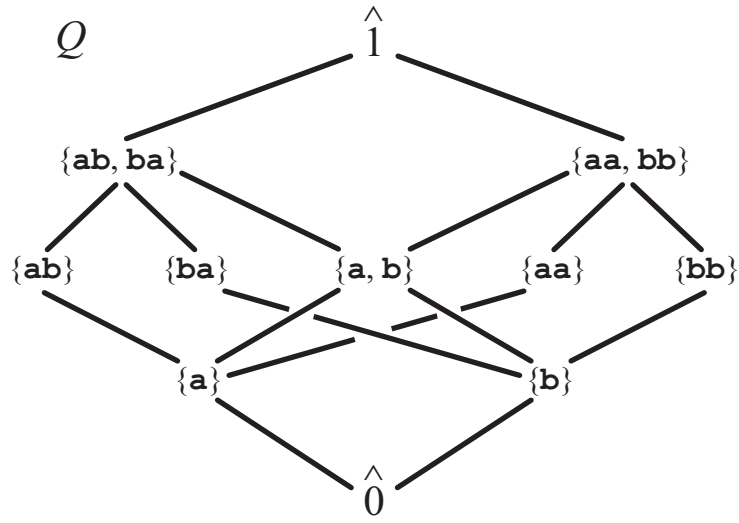


Figure 43: Poset  $Q$  modeling *sets* of attribute sequences, for inferences in the lattice of Figure 40.

produced attribute “a”. In summary, by altering the triple  $(L, P, Q)$ , one changes the possible inferences.

The poset  $Q$  of Figure 43 is a conveniently chosen finite subposet of a particular infinite poset modeling sets of sequences. In that model, each set is required to be finite and *prefix-free*, meaning that if two distinct sequences appear in an element of  $Q$ , neither may be a prefix of the other. The partial order on  $Q$  is defined by:  $q_1 \leq_Q q_2$  precisely when every sequence in  $q_1$  is a prefix of (possibly equal to) some sequence in  $q_2$ .

### 13 Lattices for Strategy Obfuscation

We have seen sublattices of powerset lattices, those being prototypical examples of Boolean lattices. A related example is given by *strategy complexes* [6, 7], which may be viewed as lattices of partial orders formed from potentially stochastic or nondeterministic transitions in a graph. The basic elements in such a lattice are *strategies* for attaining various goals. Our work on privacy now raises the question of strategy obfuscation: How can someone reveal the *actions* of a strategy in a fashion that delays identification of the strategy?

#### 13.1 Strategies for Nondeterministic Graphs

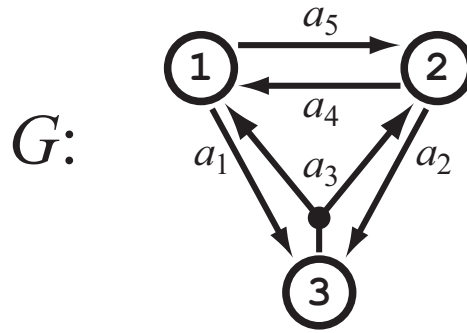


Figure 44: A graph  $G$  with three states, four deterministic actions, and one nondeterministic action ( $a_3$ ).

For a very simple example, consider the graph of Figure 44. We think of this graph as modeling some kind of dynamic system, for instance, a person driving between three shopping malls or a robot moving among clutter in a warehouse or an intruder in a server network.

There are three states in the graph, along with five actions. Each action has a *source* state and one or more *target* states. An action may be *executed* when the system is at the source of the action, causing the system to move from the action's source to one its target states.

Four of the actions,  $\{a_1, a_2, a_4, a_5\}$ , are standard deterministic directed edges, leading for certain from one state to another. The remaining action,  $a_3$ , is nondeterministic. Nondeterminism of  $a_3$  means that if the system is at state 3 and executes action  $a_3$ , then the precise outcome is uncertain: The system might move either to state 1 or to state 2. Nondeterminism is potentially adversarial: The precise target state attained is unpredictable and could vary nonstochastically on different executions of the action. One may generalize this idea to include stochastic actions along with deterministic and nondeterministic actions, thus modeling adversarial combinations of Markov chains. We will not do so here, but see [6, 7].

For our purposes here, a *strategy* is a set of actions whose underlying edge set contains no cycles. If the system is at a state which is the source of an action in the strategy, then the system executes that action. If the strategy contains multiple actions with that same source state, then the actual action executed is determined nondeterministically. For instance, in the example, if actions  $a_1$  and  $a_5$  both appear in a strategy then the strategy is agnostic as to whether the system will transition to state 2 or state 3 from state 1. If a strategy does not contain an action for a given state, then the system will stop moving if it is ever in that state.

The lattice operations for strategies are set union and set intersection, with one proviso: Suppose  $\sigma_1$  and  $\sigma_2$  are two strategies. Each strategy is a set of actions with no cycles in its underlying edge set. If the union of two strategies  $\sigma_1 \cup \sigma_2$  contains an underlying cycle in its edge set, then the lattice operation becomes  $\sigma_1 \vee \sigma_2 = \hat{1}$ , with  $\hat{1}$  the top element of the lattice. That top element represents cyclicity. The bottom element  $\hat{0}$  of the lattice is equivalent to the empty strategy  $\emptyset$ , amounting to no motion.

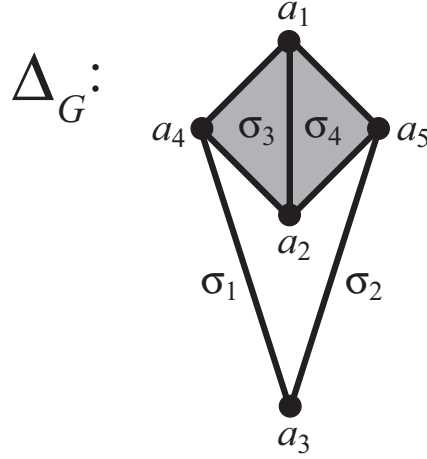


Figure 45: The strategy complex for the graph of Figure 44. We have labeled each maximal simplex with an identifier, for the purposes of Figure 46.

Rather than draw a lattice of strategies  $L$ , it is more convenient to draw an equivalent simplicial complex whose vertices are the (acyclic) actions  $\mathfrak{A}$  of the graph. This simplicial complex is denoted by  $\Delta_G$  and is called the *strategy complex* of  $G$ . The connection is that the proper part of the lattice is the face poset of the simplicial complex, that is  $L \setminus \{\hat{0}, \hat{1}\} = \mathfrak{F}(\Delta_G)$ . Figure 45 shows the strategy complex for the graph of Figure 44. The constituent simplices of the strategy complex are strategies, that is, all sets of actions whose underlying edge sets are acyclic.

$A$	$a_1$	$a_2$	$a_3$	$a_4$	$a_5$	Goal
$\sigma_1$			•	•		1
$\sigma_2$			•		•	2
$\sigma_3$	•	•		•		3
$\sigma_4$	•	•			•	3

Figure 46: Relation  $A$  describes the strategy complex of Figure 45 in terms of its maximal simplices and their constituent actions. The rightmost column shows each maximal strategy's goal, i.e., that state at which motion ceases.

Now that we have a simplicial complex, we can form a relation, whose “individuals” are all maximal strategies of the complex and whose “attributes” are the underlying actions, as shown in Figure 46. The figure also shows each maximal strategy's *goal*, that is, the state at

which the strategy would stop moving. (In general, a strategy, even a maximal strategy, may have a multi-state goal set, but in this example the goals of all maximal strategies are singleton states.) We make the following observations:

- There is at least one strategy for attaining each state in the graph, meaning it is possible to move from every state to every other state, despite uncertainty in the outcome of one of the actions. Such graphs are called *fully controllable* in [6, 7], and have properties similar to those of strongly connected directed graphs.
- For each maximal strategy, there are *two* informative attribute release sequences, each consisting of two actions. For instance, for  $\sigma_2$  one could reveal actions  $a_3$  and  $a_4$  in either order, identifying  $\sigma_2$  only after revealing both actions. For  $\sigma_3$ , one could reveal actions  $a_1$  and  $a_4$  in either order, now identifying  $\sigma_3$  only after revealing both actions.
- Some actions reveal the goal even though they do not identify the maximal strategy. In particular, actions  $a_1$  and  $a_2$  each individually reveal the goal to be 3. (The two actions are in fact equivalent in  $A$ , in that either one implies the other.) For instance, if one knows that  $a_1$  is in a maximal strategy  $\sigma$ , then one knows that the strategy cannot also contain  $a_3$ , as adding  $a_3$  would create a cycle in the underlying edge set. Action  $a_2$  must therefore also be in the strategy, since the strategy is maximal. Consequently, the goal is state 3 and  $\sigma$  is either  $\sigma_3$  or  $\sigma_4$ . The difference between these two maximal strategies is a choice between  $a_4$  and  $a_5$ . That choice does not affect the final goal, but could affect intermediate motions and the time to reach the goal. A rough analogy is knowing that a car on a freeway must continue on the freeway until at least the next exit but has a choice between lanes enroute.
- Each strategy has at least *one* informative attribute release sequence, consisting of two actions, that does *not* reveal the goal until the final action has been released. For instance, for  $\sigma_3$ , one could first release  $a_4$ , leaving open the possibility of either state 1 or state 3 being the goal, then subsequently release either  $a_1$  or  $a_2$ .

**Question:** Is this set of intertwined observations fundamental?

**Answer:** Yes, with certain qualifications, described next.

## 13.2 Connecting the Topologies of Strategy Complexes and Privacy

**Notation:**

- $G = (V, \mathfrak{A})$  denotes a graph with underlying states  $V$  and possibly uncertain actions  $\mathfrak{A}$ . (For simplicity, we assume here that both  $V$  and  $\mathfrak{A}$  are not empty.)
- $\Delta_G$  denotes the strategy complex of  $G$ ; it includes the empty strategy  $\emptyset$ .

**Lemma 29.** *Let  $G = (V, \mathfrak{A})$  be a graph as above and  $\mathfrak{M}$  the set of maximal simplices of  $\Delta_G$ .*

*Define relation  $A$  on  $\mathfrak{M} \times \mathfrak{A}$  by  $A = \{(\sigma, a) \mid a \in \sigma \in \mathfrak{M}\}$ . Then  $\Phi_A = \Delta_G$ . In other words, the Dowker complex over the set of actions is the same as the graph's strategy complex.*

(The lemma holds more generally for simplicial complexes. The proof is nearly definitional.)  
 (The “A” stands for “Action”).)

One of the fundamental results from [6, 7] is that a graph is fully controllable if and only if its strategy complex is homotopic to a sphere of dimension two less than the number of states in the graph: (Recall that “ $\simeq$ ” means homotopy equivalence.)

**Theorem 30.** *A graph  $G = (V, \mathfrak{A})$  is fully controllable if and only if  $\Delta_G \simeq \mathbb{S}^{n-2}$ , with  $n = |V|$ .*

Now recall our fundamental privacy result, Corollary 26 from page 48. That corollary, along with Theorem 30, tells us that if a graph  $G = (V, \mathfrak{A})$  is fully controllable, then the poset  $P_A$  formed from the relation  $A$  constructed as above must contain at least  $n!$  maximal chains, each consisting of at least  $n - 1$  elements, with  $n = |V|$  (recall that the number of elements in a chain is one more than its length).

We actually want a stronger result, speaking to individual strategies and we can get that by looking into the details of the proof of Theorem 25. The proof is an induction that recursively considers links, giving us the following (see Appendices G and H):

**Theorem 31** (Delaying Strategy Identification). *Let  $G = (V, \mathfrak{A})$  be a fully controllable graph, with  $n = |V| > 1$ . Let  $A$  be the relation constructed as in Lemma 29 and let  $P_A$  be its associated doubly-labeled poset. Then:*

*For each  $v \in V$ , there exists a maximal strategy  $\sigma_v \in \Delta_G$  for attaining singleton goal state  $v$  such that  $P_A$  contains at least  $(n - 1)!$  distinct maximal chains for identifying  $\sigma_v$ , with each chain consisting of at least  $n - 1$  elements.*

**Clarifying Observation:** Each maximal chain for identifying  $\sigma_v$  specifies at least  $n - 1$  actions and an order for releasing them such that no action is implied by those previously released. In particular, the sequence of actions does not identify  $\sigma_v$  until all actions have been released.

**Comments:** Theorem 31 does *not* assert that *every* maximal strategy in  $\Delta_G$  has  $(n - 1)!$  many “long” identifying chains, merely that for every possible singleton goal  $v$ , there is *some* strategy for attaining  $v$  with  $(n - 1)!$  many “long” identifying chains. It is not hard to construct an example for which some maximal strategy with a singleton goal state has fewer than  $(n - 1)!$  identifying chains. This fact suggests further questions. Here are two:

- Given an arbitrary maximal strategy  $\sigma_v$  for attaining a singleton goal state  $v$ , can we find at least *one* chain in  $P_A$  that identifies  $\sigma_v$  but requires release of at least  $n - 1$  actions before doing so? We do not know the answer in general, although small examples suggest the answer is “yes”. We do have a proof that the answer is “yes” when the graph contains a Hamiltonian cycle consisting of edges that come from deterministic or stochastic actions.
- Given a singleton goal state  $v$ , can we find at least one maximal strategy  $\sigma_v$  and at least one chain in  $P_A$  that eventually identifies  $\sigma_v$ , but does not reveal the goal  $v$  before releasing at least  $n - 1$  actions? The answer to this question is “yes”. The proof operates by repeatedly creating quotient graphs. In forming a quotient graph, the proof regards



as equivalent a certain set of states that are connected by a cycle of edges, with each edge coming from some deterministic or stochastic action. For instance, in the graph of Figure 44, the proof would regard states 1 and 2 as equivalent. The resulting quotient graph would then consist of two states with deterministic actions between them, since action  $a_3$  becomes a deterministic transition in the quotient graph. Inductively, one therefore sees that an entity can hide its true goal until at least two actions in the original graph  $G$  have been revealed. (See Appendix H for further details.)

**A comment/caution regarding the availability of many chains:** The  $(n-1)!$  chains mentioned above may come from all possible permutations of the same underlying set of  $n-1$  actions. Alternatively, these  $(n-1)!$  chains may involve creative sequencing of more than  $n-1$  actions. The precise makeup of the chains depends on the underlying homology generators. However, even if the chains are merely reordering the same  $n-1$  actions, there is good reason to take advantage of that capability, rather than pick one particular sequence via a deterministic algorithm. The reason is that knowledge of how an algorithm releases actions may leak information to an adversary. Such leakage may be understood as changing the effective relation. For instance, despite thinking one is working with relation  $A$ , a particular release protocol may simply be focusing on some proper subset of  $A$  or some proper subset of the poset  $P_A$ , possibly resulting in very different inference characteristics. A good release strategy may be to choose randomly from among the  $(n-1)!$  possible chains. In that way, one is taking good advantage of the spherical homogeneity suggested by homology.

## 14 Relations as a Category

We have discussed disinformation, obfuscation, and other manipulation of relations. The goal of such transformations has been to preserve privacy by removing free faces. We have not yet discussed such transformations formally. For instance, the coordinate transformations of Section 9 raise the question:

How should one think about maps between relations?

### 14.1 Relationship-Preserving Morphisms

Traditionally, relations are morphisms between sets (with functions a special case). In thinking about privacy, it is useful to define a category in which the relations are the objects. We have some choices in defining morphisms for this category. Bearing in mind our Dowker constructions, we make the following definitions.

**Notation:** We frequently will be working with two relations:  $R$  is a relation on  $X^R \times Y^R$  and  $Q$  is a relation on  $X^Q \times Y^Q$  (the superscripts are just indices to indicate the underlying relation). In order to distinguish rows and columns between the two, we will also use notation of the form  $X_y^R$ ,  $Y_x^R$ ,  $X_y^Q$ , and  $Y_x^Q$ .

**Definition 32** (Morphism). *Let  $R$  be a relation on  $X^R \times Y^R$  and let  $Q$  be a relation on  $X^Q \times Y^Q$ . A morphism of relations  $f : R \rightarrow Q$  is a pair of set functions:*

$$\begin{aligned} f_X &: X^R \rightarrow X^Q \\ f_Y &: Y^R \rightarrow Y^Q \end{aligned}$$

*such that  $(f_X(x), f_Y(y)) \in Q$  whenever  $(x, y) \in R$ .*

In other words, a morphism of relations maps individuals to individuals and attributes to attributes in a way that preserves relationships.

The following lemma follows from the definitions (a proof appears in Appendix I):

**Lemma 33** (Induced Simplicial Maps). *A morphism  $f : R \rightarrow Q$  between nonvoid relations induces simplicial maps between the Dowker complexes:*

$$\begin{aligned} f_X &: \Psi_R \rightarrow \Psi_Q \\ f_Y &: \Phi_R \rightarrow \Phi_Q \end{aligned}$$

**Notational comment:** The symbols  $f_X$  and  $f_Y$  are overloaded intentionally. The simplicial map  $f_X$  is precisely the set map  $f_X$  applied to the vertices of any simplex: If  $\sigma = \{x_0, \dots, x_k\} \in \Psi_R$ , then  $f_X(\sigma) = \{f_X(x_0), \dots, f_X(x_k)\} \in \Psi_Q$ . Similarly for  $f_Y$ .

Intuitively, one cannot partition the individuals of a connected relation into two or more classes without misclassifying or ignoring at least some relationships. A graph connectivity argument provides a possible proof. Lemma 33 provides another, with additional insight. Let's look at some examples:

**Two Bits onto One:** Consider again the relations  $S$  and  $Q$  of Figures 15 and 16, respectively, on page 30. Relation  $S$  models a one-bit relation — an attribute and its negation. Relation  $Q$  models a two-bit relation — two attributes and their negations. The Dowker complexes for  $S$  have  $\mathbb{S}^0$  homotopy type, while those for  $Q$  have  $\mathbb{S}^1$  homotopy type. We can think of  $S$  as a classification, splitting individuals into those that have some attribute  $\mathbf{a}$  and those that do not.

By Lemma 33, a morphism  $f : Q \rightarrow S$  induces simplicial (hence continuous) maps between the corresponding Dowker complexes of  $S$  and  $Q$ . Since  $\mathbb{S}^1$  is connected but  $\mathbb{S}^0$  is not, there is no surjective continuous function from  $\mathbb{S}^1$  to  $\mathbb{S}^0$ . Consequently, no morphism  $f : Q \rightarrow S$  can truly be a classification:  $f_Y$  can map all four attributes  $\{\mathbf{a}, \neg\mathbf{a}, \mathbf{b}, \neg\mathbf{b}\}$  of  $Q$  to the single attribute  $\mathbf{a}$  or all four attributes to  $\neg\mathbf{a}$ , but  $f_Y$  cannot map to both  $\mathbf{a}$  and  $\neg\mathbf{a}$ .

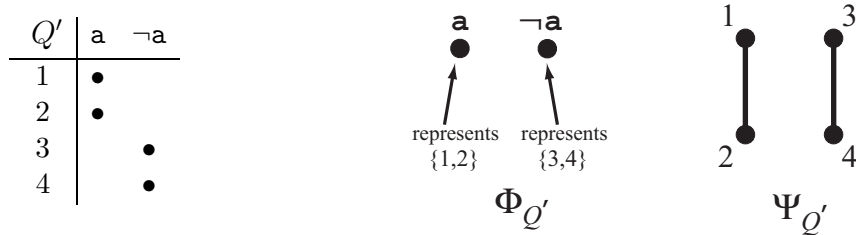


Figure 47: Relation  $Q'$  obtained from relation  $Q$  of Fig. 16 by discarding attributes  $\mathbf{b}$  and  $\neg\mathbf{b}$ .

This impossibility may at first seem paradoxical. After all, one can simply cut relation  $Q$  down the middle and throw away the columns involving attributes  $\mathbf{b}$  and  $\neg\mathbf{b}$ , as shown in Figure 47. After that, a surjective morphism  $f' : Q' \rightarrow S$  is immediate. Indeed, that is possible. However, in so doing, one has discarded some relationships, perhaps purposefully, perhaps accidentally. In particular, the relationship between individuals #1 and #3 of  $Q$  via attribute  $\mathbf{b}$  is lost, as is the relationship between individuals #2 and #4 via attribute  $\neg\mathbf{b}$ . This reasoning simply underscores the fact that morphisms of relations preserve relationships. Lack of continuity in a function therefore is a sign that one is discarding some relationships. Whether such discard is desirable depends on one's goals in a particular application.

**Three Bits onto Two:** Recall as well Figure 17, which depicts a three-bit relation  $R$  — three attributes and their negations, capable of distinguishing between eight individuals. The homotopy type of the Dowker complexes is  $\mathbb{S}^2$ . With  $Q$  as above, the following question arises naturally when trying to reduce complexity of data yet preserve information:

Does there exist a surjective morphism  $f : R \rightarrow Q$  ?

Unlike the previous example, there do exist continuous maps from  $\mathbb{S}^2$  onto  $\mathbb{S}^1$ , so perhaps one can find a surjective morphism  $f : R \rightarrow Q$ . In fact, one cannot, for dimensional reasons that force an equator of  $\mathbb{S}^2$  to become a homology generator of  $\mathbb{S}^1$ . A simplex-based argument goes as follows:

- Suppose surjective  $f : R \rightarrow Q$  exists. As will be discussed later (see page 74), this means the component functions  $f_X : \Psi_R \rightarrow \Psi_Q$  and  $f_Y : \Phi_R \rightarrow \Phi_Q$  are surjective as set maps.

- One may therefore assume without loss of generality that  $f_Y(\mathbf{a}) = \mathbf{a}$  and  $f_Y(\mathbf{b}) = \mathbf{b}$ .
- The triangles  $\{\mathbf{a}, \mathbf{b}, \mathbf{c}\}$  and  $\{\mathbf{a}, \mathbf{b}, \neg\mathbf{c}\}$  are both simplices in  $\Phi_R$ . The maximal simplices of  $\Phi_Q$  are edges.
- By Lemma 33, this means that  $f_Y(\mathbf{c})$  and  $f_Y(\neg\mathbf{c})$  are both elements of  $\{\mathbf{a}, \mathbf{b}\}$  in  $\Phi_Q$ .
- Again by surjectivity, we therefore see that  $\{f_Y(\neg\mathbf{a}), f_Y(\neg\mathbf{b})\} = \{\neg\mathbf{a}, \neg\mathbf{b}\}$ .
- Another triangle-versus-edge argument then says that  $f_Y(\mathbf{c})$  and  $f_Y(\neg\mathbf{c})$  are both elements of  $\{\neg\mathbf{a}, \neg\mathbf{b}\}$ , giving us a contradiction.

Of course, as in constructing  $Q'$  of Figure 47, if we are willing to tolerate discontinuities, we could discard one attribute and its negation to obtain  $Q$  from  $R$ . As before, discontinuity means losing awareness of some relationship(s). For instance, if we omit attribute  $\mathbf{c}$ , we would become unaware in  $Q$  of the relationship that exists in  $R$  among the set of individuals  $\{1, 3, 5, 7\}$ .

## 14.2 Privacy-Establishing Morphisms

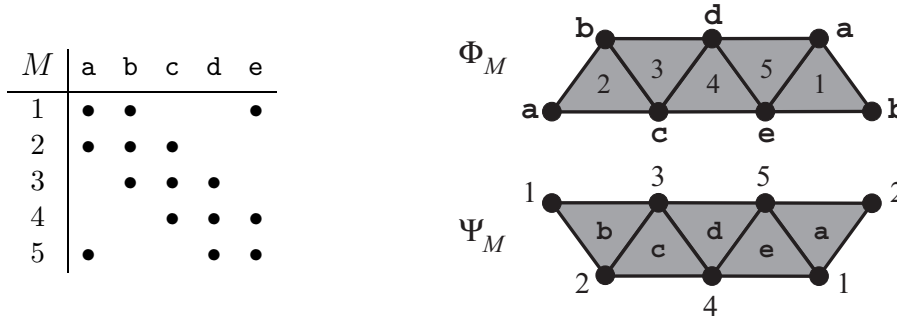


Figure 48: Relation  $M$  is isomorphic to relation  $G$  of Figure 23 on page 37, now without the author-book semantics. The Dowker complexes are dual triangulations of the Möbius strip, with  $\mathbb{S}^1$  homotopy type.

Relations involve two spaces. Looking at just  $\Phi_R$  or just  $\Psi_R$  may hide some interesting properties. For instance, consider the Möbius strip relation  $M$  of Figure 48. We encountered this relation previously, in Section 10.

We might want to try to remove some of the inferences discussed in Section 10 by reshaping the underlying relation without discarding any relationships. Doing so leads to the following question:

Does there exist a surjective morphism  $f : M \rightarrow T$ , with  $T$  a relation that preserves attribute and association privacy ?

In Section 8, we mentioned that any such  $T$  must have the topology of either a linear cycle or a spherical boundary complex. It turns out that the answer to this question is “yes” and that the relevant  $T$  creates Dowker complexes that are boundaries of tetrahedra (see Figure 30 on page 45).

This construction is not immediately obvious from the complexes  $\Phi_M$  and  $\Psi_M$ . Although those simplicial complexes are 2-dimensional, suggesting that their triangles can be wrapped around a tetrahedron, doing so actually collapses two of the five triangles to edges. Indeed, the component functions for one such surjective morphism  $f : M \rightarrow T$  are:

$$\begin{array}{ll}
 f_X : X^M \rightarrow X^T & f_Y : Y^M \rightarrow Y^T \\
 1 \mapsto 4 & \mathbf{a} \mapsto \mathbf{a} \\
 2 \mapsto 1 & \mathbf{b} \mapsto \mathbf{b} \\
 3 \mapsto 2 & \mathbf{c} \mapsto \mathbf{c} \\
 4 \mapsto 3 & \mathbf{d} \mapsto \mathbf{d} \\
 5 \mapsto 4 & \mathbf{e} \mapsto \mathbf{a}
 \end{array}$$

The induced simplicial maps act on the five maximal simplices of  $\Psi_M$  and  $\Phi_M$  as follows:

$$\begin{array}{ll}
 f_X : \Psi_M \rightarrow \Psi_T & f_Y : \Phi_M \rightarrow \Phi_T \\
 \{1, 2, 3\} \mapsto \{1, 2, 4\} & \{\mathbf{a}, \mathbf{b}, \mathbf{c}\} \mapsto \{\mathbf{a}, \mathbf{b}, \mathbf{c}\} \\
 \{2, 3, 4\} \mapsto \{1, 2, 3\} & \{\mathbf{b}, \mathbf{c}, \mathbf{d}\} \mapsto \{\mathbf{b}, \mathbf{c}, \mathbf{d}\} \\
 \{3, 4, 5\} \mapsto \{2, 3, 4\} & \{\mathbf{c}, \mathbf{d}, \mathbf{e}\} \mapsto \{\mathbf{a}, \mathbf{c}, \mathbf{d}\} \\
 \{1, 4, 5\} \mapsto \{3, 4\} & \{\mathbf{a}, \mathbf{d}, \mathbf{e}\} \mapsto \{\mathbf{a}, \mathbf{d}\} \\
 \{1, 2, 5\} \mapsto \{1, 4\} & \{\mathbf{a}, \mathbf{b}, \mathbf{e}\} \mapsto \{\mathbf{a}, \mathbf{b}\}
 \end{array}$$

Even though  $f_X$  and  $f_Y$  are surjective as set maps on the vertices of the Dowker complexes, they are *not* surjective as simplicial maps on the complexes themselves. Each only covers 3 of the 4 triangles comprising its image tetrahedron. At first glance it may therefore seem that the morphism  $f : M \rightarrow T$  resulting from  $f_X$  and  $f_Y$  does not achieve the desired privacy preservation. A closer look, however, reveals that  $f$  is actually surjective as a map of relations: it maps all the elements of  $M$  *onto* the elements of  $T$ . Therefore, it does represent a transformation that achieves privacy preservation.

In order to understand this paradox, imagine again that  $M$  represents an authorship database. Think of the maps  $f_X$  and  $f_Y$  as quotient maps, in this case equating authors 1 and 5 and books **a** and **e**. The equivalencing of authors might constitute a recognition of pseudonyms. The equivalencing of books might represent a generalization from titles to genres. Such changes of resolution, carefully chosen, perhaps based on external structure, can preserve relationships while reducing recognition and inference granularity.

### 14.3 Summary of Morphism Properties

Definition 32 defines a morphism of relations  $f : R \rightarrow Q$  in terms of underlying set functions  $f_X : X^R \rightarrow X^Q$  and  $f_Y : Y^R \rightarrow Y^Q$ . These set functions further induce simplicial maps  $f_X : \Psi_R \rightarrow \Psi_Q$  and  $f_Y : \Phi_R \rightarrow \Phi_Q$ . The previous subsections spoke of surjectivity in varying contexts. Similarly, one could speak of maps as being one-to-one in varying contexts. Finally,

one also speaks of morphisms as being epimorphisms and monomorphisms. This subsection summarizes how these properties relate for the various maps. See Appendix I for proofs.

First, some definitional context and reminders:

- Suppose  $f : R \rightarrow Q$  is a morphism of relations. Recall from category theory that  $f$  is an *epimorphism* if, for any pair of morphisms  $g, h : Q \rightarrow S$ ,  $g \circ f = h \circ f$  implies  $g = h$ .

Recall further that a morphism  $f : R \rightarrow Q$  is a *monomorphism* if, for any pair of morphisms  $g, h : S \rightarrow R$ ,  $f \circ g = f \circ h$  implies  $g = h$ .

- A morphism of relations  $f : R \rightarrow Q$  is also a set map between the set of pairs comprising  $R$  and the set of pairs comprising  $Q$ . Specifically,  $f(x, y) = (f_X(x), f_Y(y))$ .

One may speak of  $f$  as being *surjective* and/or *one-to-one*, meaning as a set map.

- The functions  $f_X : X^R \rightarrow X^Q$  and  $f_Y : Y^R \rightarrow Y^Q$  are set maps. One may speak of them as being surjective and/or one-to-one.
- One may also ask whether the induced simplicial maps  $f_X : \Psi_R \rightarrow \Psi_Q$  and  $f_Y : \Phi_R \rightarrow \Phi_Q$  are surjective and/or injective as maps between simplicial complexes viewed as sets.

**Lemma 34** (Morphism Properties). *Assume the notation from above and that all relevant relations are nonvoid. Let  $f : R \rightarrow Q$  be a morphism of relations (as per Definition 32). Then:*

(i)  $f_X$  and  $f_Y$  are one-to-one set maps  $\implies f$  is one-to-one  $\iff f$  is a monomorphism.

(ii)  $f$  surjective  $\implies f$  epimorphism  $\iff f_X$  and  $f_Y$  are surjective set maps.

(Additional conditions for that last  $\iff$ : The  $\implies$  direction assumes that  $Q$  has no blank rows or columns, while the  $\impliedby$  direction assumes that  $R$  has no blank rows or columns.)

The two uni-directional implications  $\implies$  above are strict.

(iii) If  $f_X : \Psi_R \rightarrow \Psi_Q$  is surjective and  $Q$  has no blank rows, then  $f_X : X^R \rightarrow X^Q$  is surjective.

Similarly for  $f_Y$ , now assuming that  $Q$  has no blank columns.

The converses need not hold. Indeed,  $f$  itself can be surjective but the maps of simplicial complexes need not be (as we saw with the maps of page 73).

(iv) If  $f_X : X^R \rightarrow X^Q$  is one-to-one, then  $f_X : \Psi_R \rightarrow \Psi_Q$  is injective. The converse holds if  $R$  has no blank rows.

Similarly for  $f_Y$ , now assuming that  $R$  has no blank columns for the converse.

#### 14.4 G-morphisms

Since a relation  $R$  defines a poset  $P_R$ , rather than merely create morphisms from set maps between individuals and attributes as in Definition 32, we may broaden the definition by considering maps between posets:

**Definition 35** (G-Morphism). *Let  $R$  and  $Q$  be relations.*

A G-morphism  $f : R \rightarrow Q$  is any poset map  $f : P_R \rightarrow P_Q$ .

**Comments:** The “G” stands for “Galois”. We might have insisted that a G-morphism  $R \rightarrow Q$  be a lattice morphism  $P_R^+ \rightarrow P_Q^+$  rather than merely a poset map  $P_R \rightarrow P_Q$ , but that might be too restrictive. Instead, as subsequent lemmas will describe, we view a G-morphism as providing homotopy flexibility. In particular, a morphism between relations as per Definition 32 induces two homotopic G-morphisms. The lattice structure of the image is relevant in that it allows one to fill in elements not directly in the image of any one poset map, as will become apparent in Theorem 40.

$$\begin{array}{ccc}
 \mathfrak{F}(\Psi_R) & \xrightarrow{f_X} & \mathfrak{F}(\Psi_Q) \\
 \phi_R \downarrow \uparrow \psi_R & & \phi_Q \downarrow \uparrow \psi_Q \\
 \mathfrak{F}(\Phi_R) & \xrightarrow{f_Y} & \mathfrak{F}(\Phi_Q)
 \end{array}$$

Figure 49: Diagram showing the poset maps  $f_X$  and  $f_Y$  induced by a morphism  $f : R \rightarrow Q$ , along with the homotopy equivalences between each relation’s face posets. (The diagram need not be commutative, but is almost so; see Lemma 36.)

Recall that a morphism  $f : R \rightarrow Q$  as per Definition 32 is built from two set maps  $f_X$  and  $f_Y$  and that these set maps induce simplicial maps between the Dowker complexes, as per Lemma 33. We may therefore further regard  $f_X$  and  $f_Y$  as poset maps between the face posets of the Dowker complexes:  $f_X : \mathfrak{F}(\Psi_R) \rightarrow \mathfrak{F}(\Psi_Q)$  and  $f_Y : \mathfrak{F}(\Phi_R) \rightarrow \mathfrak{F}(\Phi_Q)$ . Consequently we have a diagram of maps as in Figure 49. The diagram need not be commutative, but the following containments hold:

**Lemma 36** (Containment). *Let  $f : R \rightarrow Q$  be a morphism of nonvoid relations. Then:*

- (a)  $(f_Y \circ \phi_R)(\sigma) \subseteq (\phi_Q \circ f_X)(\sigma)$ , for every  $\sigma \in \Psi_R$ ,
- (b)  $(f_X \circ \psi_R)(\gamma) \subseteq (\psi_Q \circ f_Y)(\gamma)$ , for every  $\gamma \in \Phi_R$ .

As a corollary, we see that the diagram of Figure 49 describes two pairs of homotopic maps:

**Corollary 37** (Homotopic Face Maps). *Let  $f : R \rightarrow Q$  be a morphism of nonvoid relations. Then:*

- (a)  $f_X$  and  $\psi_Q \circ f_Y \circ \phi_R$  are homotopic poset maps  $\mathfrak{F}(\Psi_R) \rightarrow \mathfrak{F}(\Psi_Q)$ ,
- (b)  $f_Y$  and  $\phi_Q \circ f_X \circ \psi_R$  are homotopic poset maps  $\mathfrak{F}(\Phi_R) \rightarrow \mathfrak{F}(\Phi_Q)$ .

The images of the compositions that appear in Corollary 37 may be regarded as lying in  $P_Q$ . We may further restrict the domain of these maps to be  $P_R$ , giving us the following G-morphisms:

**Definition 38** (Induced G-Morphism). *A morphism of relations  $f : R \rightarrow Q$  induces two G-morphisms  $P_R \rightarrow P_Q$ , defined as follows:*

$$f_X^g = (\psi_Q \circ f_Y \circ \phi_R)|_{P_R} \quad f_Y^g = (\phi_Q \circ f_X \circ \psi_R)|_{P_R}.$$

(The  $g$  superscript stands for “Galois” while the vertical bar  $|$  means “restricted to”. See also Appendix I.2.)

**Corollary 39** (Homotopic Poset Maps). *Let  $f : R \rightarrow Q$  be a morphism of nonvoid relations. The induced G-morphisms  $f_X^g, f_Y^g : P_R \rightarrow P_Q$  are homotopic.*

Corollary 39 says that we may view the underlying maps  $f_X$  and  $f_Y$  of a morphism  $f$  as mapping any inference-closed set (viewed either as a set of individuals or as a set of attributes) from the domain of  $f$  to an interval of inference-closed sets in the codomain of  $f$ .

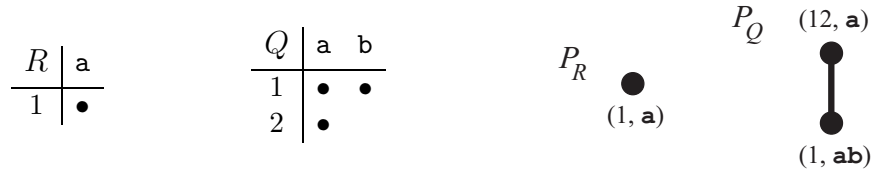


Figure 50: Relation  $R$  is a subrelation of  $Q$ . How should one embed  $P_R$  into  $P_Q$ ? There are two possible embeddings, related by a homotopy.

For a simple example, see Figure 50. One may regard relation  $R$  as a subrelation of  $Q$ , then define  $f : R \rightarrow Q$  to be inclusion. For instance, maybe  $R$  and  $Q$  represent individuals #1 and #2 at two parties  $a$  and  $b$ , with  $R$  representing known parties and party-attendees at some time and  $Q$  representing an update of that information at a later time. Observe that:

$$\begin{aligned} f_X^g((1, a)) &= (\psi_Q \circ f_Y \circ \phi_R)(\{1\}) = (\psi_Q \circ f_Y)(\{a\}) = \psi_Q(\{a\}) = \{1, 2\} \text{ “=” } (12, a), \\ f_Y^g((1, a)) &= (\phi_Q \circ f_X \circ \psi_R)(\{a\}) = (\phi_Q \circ f_X)(\{1\}) = \phi_Q(\{1\}) = \{a, b\} \text{ “=” } (1, ab). \end{aligned}$$



The last equality in each row indicates how to view the image element on the left of the “=” as an element of the poset  $P_Q$ .

Both  $f_X^g$  and  $f_Y^g$  tell us how to update inference-closed sets from  $P_R$  into inference-closed sets within  $P_Q$ :

- The map  $f_X^g$  updates association inferences while holding observed attributes fixed. In this example, based on initial information (relation  $R$ ), we know that person #1 attended party **a**. Once we update that information (relation  $Q$ ) we can conclude that person #2 also attended a party at which person #1 was present.
- Similarly, the map  $f_Y^g$  updates attribute inferences while holding observed individuals fixed. In this example, updated information allows us to conclude that person #1 attended not only party **a** but also party **b**.

In general, for any fixed element of  $P_R$ , the two maps may give different results, but those results are comparable in  $P_Q$ . Here  $f$  was inclusion, so we could speak of holding attributes or individuals “fixed”. More generally, “fixed” is replaced by whatever  $f$  does.

## 14.5 Surjectivity Revisited

A paradox: We saw on page 73 a surjective morphism  $f$ , from the Möbius strip relation of Figure 48 to the tetrahedral relation of Figure 30, whose induced simplicial maps  $f_X : \Psi_M \rightarrow \Psi_T$  and  $f_Y : \Phi_M \rightarrow \Phi_T$  were not surjective. This raises some questions:

1. Are the induced poset maps  $f_X^g, f_Y^g : P_M \rightarrow P_T$  surjective?
2. If not, how can one speak of a surjective morphism?

(Note that  $P_M^+$  is isomorphic to  $P_G^+$  as shown in Figure 25 on page 39. A rendering would be identical, except for lowercase letters in place of uppercase ones. The lattice  $P_T^+$  appears in Figure 31 on page 46.)

The answer to Question 1 is that the two poset maps are *not* surjective. Observe in Table 9, for instance, that the image of  $f_X^g$  does not include  $(4, \mathbf{abd})$ . Similarly, the image of  $f_Y^g$  does not include  $(134, \mathbf{a})$ .

These missing elements *are* in the image of both maps *together*, viewed as a pair of homotopic maps, as per Corollary 39. Unfortunately, that explanation is not a full answer to Question 2. For instance, neither map’s image includes the element  $(13, \mathbf{ac})$  of  $P_T$ , nor does that element appear in any interval  $[f_Y^g(p), f_X^g(p)]$  as  $p$  varies throughout  $P_M$ .

To answer question 2, the lattice structure of  $P_T$  is useful. In the example, the image of  $f_X^g$  includes all elements of  $P_T$  that correspond to maximal simplices of  $\Psi_T$ . Similarly, the image of  $f_Y^g$  includes all elements of  $P_T$  that correspond to maximal simplices of  $\Phi_T$ . Intuitively, we therefore expect that the lattice operations (which correspond to intersection in either  $\Psi_T$  or  $\Phi_T$ ) will generate all the elements of  $P_T$ . In that sense, the surjectivity of  $f$  appears as surjectivity of each of  $f_X^g$  and  $f_Y^g$ , once one *completes* their images under lattice operations.

$p$	$f_X^g(p)$	$f_Y^g(p)$
(12 , ab)	(14 , ab)	(14 , ab)
(2 , abc)	(1 , abc)	(1 , abc)
(123 , b)	(124 , b)	(124 , b)
(23 , bc)	(12 , bc)	(12 , bc)
(3 , bcd)	(2 , bcd)	(2 , bcd)
(234 , c)	(123 , c)	(123 , c)
(34 , cd)	(23 , cd)	(23 , cd)
(4 , cde)	(3 , acd)	(3 , acd)
(345 , d)	(234 , d)	(234 , d)
(45 , de)	(34 , ad)	(34 , ad)
(5 , ade)	(34 , ad)	(4 , abd)
(145 , e)	(134 , a)	(34 , ad)
(15 , ae)	(134 , a)	(4 , abd)
(1 , abe)	(14 , ab)	(4 , abd)
(125 , a)	(134 , a)	(14 , ab)

Table 9: Each  $p$  is of the form  $(\sigma, \gamma) \in P_M$ . The elements  $f_X^g(p)$  and  $f_Y^g(p)$  lie in  $P_T$ . See also Figures 25 and 31, on pages 39 and 46, respectively. (As in those figures, the table elides commas and braces from set notation.)

The following theorem summarizes the intuition of the previous pages:

**Theorem 40** (Lattice Surjectivity). *Let  $R$  and  $Q$  be nonvoid relations with no blank rows or columns. Suppose  $f : R \rightarrow Q$  is a surjective morphism (in the sense of Definition 32). For any  $q \in P_Q$ :*

$$q = \bigwedge_j \bigvee_i q_{ji}, \quad \text{with each } q_{ji} \text{ in the image of } f_X^g : P_R \rightarrow P_Q,$$

$$q = \bigvee_k \bigwedge_\ell q'_{k\ell}, \quad \text{with each } q'_{k\ell} \text{ in the image of } f_Y^g : P_R \rightarrow P_Q.$$

## Acknowledgments

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We are also grateful for the publicly available **Perseus** software previously written at the University of Pennsylvania, which we used for our homology computations: <http://www.sas.upenn.edu/~vnanda/perseus/>.

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## A Preliminaries

**Assumption:** All simplicial complexes, relations, posets, and lattices in this document are finite.

### A.1 Simplicial Complexes

We largely follow the notation of [14] and [1].

- An *(abstract) simplicial complex*  $\Sigma$  with underlying vertex set  $X$  is a collection of finite subsets of  $X$ , such that if  $\sigma$  is in  $\Sigma$  then so is every subset of  $\sigma$ . The elements of  $\Sigma$  are *simplices*. We allow the empty set  $\emptyset$  to be a simplex in  $\Sigma$ , for combinatorial reasons. We refer both to the elements of a simplex and to singleton simplices as *vertices*. Not all elements of  $X$  need to be vertices of  $\Sigma$ .
- We let  $\text{verts}(\Sigma)$  denote the set of vertices that actually appear in  $\Sigma$ , called the *zero-skeleton* of  $\Sigma$ . [The standard notation is  $\Sigma^{(0)}$  but that conflicts with some iterative notation in the proof of Theorem 25.]
- The *dimension* of a simplex  $\sigma$  is one less than its cardinality. The empty simplex  $\emptyset$  has dimension  $-1$ . If a simplex has dimension  $k$  we sometimes call it a *k-simplex*.
- The *void complex*  $\emptyset$  has no simplices in it. The void complex is degenerate. The *empty complex*  $\{\emptyset\}$  consists solely of the empty simplex. The empty complex represents the empty topological space. It is also the sphere of dimension  $-1$ , written  $\mathbb{S}^{-1}$ . (There could be different instances of the void or empty complex, depending on the underlying vertex set  $X$ , though frequently one takes that to be empty.)
- A simplex  $\sigma$  of a simplicial complex  $\Sigma$  is a *free face* of  $\Sigma$  if it is a proper subset of exactly one maximal simplex  $\tau$  of  $\Sigma$ . (The empty simplex  $\emptyset$  can sometimes be a free face.)
- With  $\Sigma$  a simplicial complex,  $C_k(\Sigma; \mathbb{Z})$  is the group of simplicial  $k$ -chains over  $\Sigma$  with integer coefficients. A  $k$ -chain  $c \in C_k(\Sigma; \mathbb{Z})$  assigns to each oriented  $k$ -simplex  $\tau$  an integer, such that  $c(-\tau) = -c(\tau)$ . (Caution: We will later use the word “chain” in the poset sense; there should be no ambiguity given context.)
- Suppose  $\Sigma$  is a simplicial complex and  $c \in C_k(\Sigma; \mathbb{Z})$ . Assume all simplices have been assigned an orientation in  $\Sigma$ . One can write  $c = \sum_i n_i \tau_i$  uniquely, for some collection of (oriented)  $k$ -dimensional simplices  $\{\tau_i\}$  in  $\Sigma$  such that  $n_i \neq 0$  for each  $i$ . This means  $c(\tau_i) = n_i$  for each  $\tau_i$  that appears in the sum and  $c(\tau) = 0$  for all other  $k$ -simplices  $\tau$ .  
We define the *support* of  $c$  as  $\|c\| = \cup_i \tau_i$ . The support is the set of all vertices that appear in any of the simplices  $\tau$  for which  $c(\tau)$  is nonzero.
- We let  $\partial$  and  $\tilde{\partial}$  stand for “boundary”. There are two contexts:
  1. When  $V$  is a nonempty finite set of points, then  $\partial(V)$  means the simplicial complex whose underlying vertex set is  $V$  and whose simplices consist of all proper subsets

of  $V$ . We refer to this complex as the *boundary complex of the full simplex on vertex set  $V$* . It has the homotopy type of a sphere, specifically  $\mathbb{S}^{n-2}$ , with  $n = |V|$ , for all  $n \geq 1$ .

2. We also designate the *simplicial boundary operator* by  $\partial$  and the *reduced boundary operator* by  $\tilde{\partial}$ . These operators are families of maps, describing for each dimension  $k$  a group homomorphism  $C_k(\Sigma; \mathbb{Z}) \rightarrow C_{k-1}(\Sigma; \mathbb{Z})$ .

(See below for the special case  $k = 0$ .)

Given an oriented  $k$ -simplex  $\sigma = \{x_0, \dots, x_k\}$ , with  $k \geq 1$ ,  $\tilde{\partial}_k(\sigma) = \partial_k(\sigma) = \sum_{i=0}^k (-1)^i \tau_i$ , where  $\tau_i$  is the oriented  $(k-1)$ -simplex formed from  $\sigma$  by removing vertex  $x_i$  and using the induced orientation of  $\sigma$  on  $\tau_i$ .

For  $k = 0$ ,  $\partial_0 : C_0(\Sigma; \mathbb{Z}) \rightarrow 0$ , while  $\tilde{\partial}_0 : C_0(\Sigma; \mathbb{Z}) \rightarrow \mathbb{Z}$ , with  $\tilde{\partial}_0(\{v\}) = 1$ , for each vertex  $\{v\} \in \Sigma$ . There is also a map  $\tilde{\partial}_{-1} : \mathbb{Z} \rightarrow 0$ . See [14, 12] for further details.

We are mainly interested in the reduced boundary operator  $\tilde{\partial}$ .

We often write  $\tilde{\partial}$  in place of  $\tilde{\partial}_k$  when the context  $k$  is clear.

Elements of the subgroup  $\ker(\tilde{\partial}_k)$  are called *reduced  $k$ -cycles*.

Elements of the subgroup  $\text{img}(\tilde{\partial}_{k+1})$  are called *reduced  $k$ -boundaries*.

- Given a simplicial complex  $\Sigma$ ,  $\tilde{H}_k(\Sigma; \mathbb{Z})$  is the *reduced homology group in dimension  $k$*  based on simplicial chains over  $\Sigma$  with integer coefficients. It is a quotient group, measuring the number of reduced  $k$ -cycles that are not reduced  $k$ -boundaries.

Formally,  $\tilde{H}_k(\Sigma; \mathbb{Z}) = \ker(\tilde{\partial}_k) / \text{img}(\tilde{\partial}_{k+1})$ . (That makes sense since  $\tilde{\partial}_k \circ \tilde{\partial}_{k+1} = 0$ .)

- Given a simplicial complex  $\Sigma$  and a set  $\sigma$ , we define the following three simplicial subcomplexes of  $\Sigma$  in the standard way:
  - The *link* of  $\sigma$  in  $\Sigma$ :  $\text{Lk}(\Sigma, \sigma) = \{\tau \in \Sigma \mid \tau \cap \sigma = \emptyset \text{ and } \tau \cup \sigma \in \Sigma\}$ .
  - The *deletion* of  $\sigma$  in  $\Sigma$ :  $\text{dl}(\Sigma, \sigma) = \{\tau \in \Sigma \mid \tau \cap \sigma = \emptyset\}$ .
  - The *closed star* of  $\sigma$  in  $\Sigma$ :  $\overline{\text{St}}(\Sigma, \sigma) = \{\tau \in \Sigma \mid \tau \cup \sigma \in \Sigma\}$ .

The definitions make sense even when  $\sigma$  is not itself a simplex in  $\Sigma$ , though in that case both  $\text{Lk}(\Sigma, \sigma)$  and  $\overline{\text{St}}(\Sigma, \sigma)$  are the void complex  $\emptyset$ .

Observe that  $\text{dl}(\Sigma, \sigma) \cap \overline{\text{St}}(\Sigma, \sigma) = \text{Lk}(\Sigma, \sigma)$  and  $\overline{\text{St}}(\Sigma, \sigma) = \text{Lk}(\Sigma, \sigma) * \langle \sigma \rangle$ .

Here  $*$  means simplicial join (described below) and  $\langle \sigma \rangle$  is the simplicial complex generated by  $\sigma$ , consisting of all subsets of  $\sigma$ .

When  $\sigma$  consists of a single element  $v$ , i.e.,  $\sigma = \{v\}$ , we tend simply to write  $\text{Lk}(\Sigma, v)$ ,  $\text{dl}(\Sigma, v)$ ,  $\overline{\text{St}}(\Sigma, v)$ . Aside: For a singleton  $v$ , it is further true that  $\text{dl}(\Sigma, v) \cup \overline{\text{St}}(\Sigma, v) = \Sigma$ .

- One may associate a *geometric realization* to a finite abstract simplicial complex  $\Sigma$  by embedding  $\Sigma$  into a finite-dimensional Euclidean space. One may therefore think of  $\Sigma$  as a topological space [14, 1].

- Suppose  $\Sigma$  and  $\Gamma$  are two simplicial complexes with underlying vertex sets  $X$  and  $Y$ , respectively. A set function  $f : X \rightarrow Y$  is said to be a *simplicial map* if it satisfies the following condition: If  $\sigma \in \Sigma$ , then  $f(\sigma) \in \Gamma$ .

In that case, one may view  $f$  as a map of simplicial complexes,  $f : \Sigma \rightarrow \Gamma$ .

A simplicial map may further be viewed as a continuous function between the geometric realizations of  $\Sigma$  and  $\Gamma$ .

- When  $X_1$  and  $X_2$  are topological spaces, the notation  $X_1 \simeq X_2$  means that  $X_1$  and  $X_2$  have the same *homotopy type* [1, 12].
- When  $X_1$  and  $X_2$  are topological spaces,  $X_1 \vee X_2$  means a *wedge sum* of  $X_1$  and  $X_2$  [12].
- Suppose  $\mathcal{U}$  is a finite nonvoid collection of (not necessarily distinct) topological subspaces of some ambient space. One may define a simplicial complex  $\mathcal{N}(\mathcal{U})$ , called the *nerve of  $\mathcal{U}$* : The simplices of  $\mathcal{N}(\mathcal{U})$  are given by the empty simplex and all nonvoid finite subcollections  $\{U_1, \dots, U_k\}$  of  $\mathcal{U}$  such that  $U_1 \cap \dots \cap U_k \neq \emptyset$ . Under certain conditions, if these nonempty intersections are contractible, then the nerve has the same homotopy type as the union of all the elements in  $\mathcal{U}$ :  $\mathcal{N}(\mathcal{U}) \simeq \bigcup_{U \in \mathcal{U}} U$ . See [1, 12] for conditions.
- Suppose  $\Sigma$  and  $\Gamma$  are simplicial complexes with disjoint underlying vertex sets. The *simplicial join* [18] of  $\Sigma$  and  $\Gamma$  is the simplicial complex

$$\Sigma * \Gamma = \{\sigma \cup \gamma \mid \sigma \in \Sigma \text{ and } \gamma \in \Gamma\}.$$

The underlying vertex set of  $\Sigma * \Gamma$  is the union of the underlying vertex sets of  $\Sigma$  and  $\Gamma$ .

## A.2 Partially Ordered Sets (Posets)

We largely follow the notation of [18].

- A *poset*  $P$  is a set of elements with a partial order, sometimes written simply as “ $\leq$ ” other times as “ $\leq_P$ ”. The symbols “ $\geq$ ”, “ $<$ ”, “ $>$ ” and “ $=$ ” are defined accordingly.
- A *chain*  $c$  in a poset  $P$  is a totally ordered subset of  $P$ , which we often write as  $c = \{p_0 < p_1 < \dots < p_\ell\}$ . The *length*  $\ell(c) = \ell$  of a chain  $c$  is one less than the number of elements in the chain (much like simplex dimension). The length of the empty chain is  $-1$ . The length  $\ell(P)$  of a poset  $P$  is the maximum length of any chain in  $P$ .
- The *face poset*  $\mathfrak{F}(\Sigma)$  of a nonvoid simplicial complex  $\Sigma$  consists of all nonempty simplices of  $\Sigma$ , partially ordered by set inclusion.
- The *order complex*  $\Delta(P)$  of a poset  $P$  is the simplicial complex whose simplices are given by all finite chains  $\{p_0 < p_1 < \dots < p_\ell\}$  of  $P$ . (If  $P = \emptyset$ , then  $\Delta(P) = \{\emptyset\}$ .)
- One may speak of the *topology of a poset*: One says that a poset  $P$  has a topological property when its order complex  $\Delta(P)$  has that property. For instance, to say that a poset is contractible means that its order complex is contractible. To say that two posets  $P$  and  $Q$  are homotopic means that  $\Delta(P)$  and  $\Delta(Q)$  have the same homotopy type. Etc.

- It is a fact that  $\Delta(\mathfrak{F}(\Sigma))$  is homeomorphic to  $\Sigma$ . Indeed,  $\Delta(\mathfrak{F}(\Sigma))$  may be viewed as the *first barycentric subdivision* of  $\Sigma$ , which we write as  $\text{sd}(\Sigma)$ . See [18, 16].
- A set function  $\theta : P \rightarrow Q$  between two posets  $P$  and  $Q$  is said to be a *poset map* if it is either *order-preserving* or *order-reversing*. That means, for all  $x, y \in P$ :

$$\begin{aligned} \text{order-preserving:} \quad & \text{If } x \leq_P y, \text{ then } \theta(x) \leq_Q \theta(y). \\ \text{order-reversing:} \quad & \text{If } x \leq_P y, \text{ then } \theta(x) \geq_Q \theta(y). \end{aligned}$$

- A poset map  $\theta : P \rightarrow Q$  between two posets  $P$  and  $Q$  induces a simplicial map between the associated order complexes  $\theta : \Delta(P) \rightarrow \Delta(Q)$ .
- An order-preserving poset self-map  $\theta : P \rightarrow P$  is said to be a *closure operator* when  $x \leq \theta(x)$ , for all  $x \in P$ , and  $\theta \circ \theta = \theta$ . A closure operator  $\theta$  defines a homotopy equivalence between  $P$  and the image  $\theta(P)$ . See [1, 18].

### A.3 Semi-Lattices and Lattices

Let  $L$  be a partially ordered set and suppose  $p, q \in L$ :

- If  $p$  and  $q$  have a least upper bound, then one writes  $p \vee q$  to mean that least upper bound. If every pair of elements has a least upper bound, one says that  $L$  is a *join semi-lattice*.
- If  $p$  and  $q$  have a greatest lower bound, then one writes  $p \wedge q$  to mean that greatest lower bound. If every pair of elements has a greatest lower bound, one says that  $L$  is a *meet semi-lattice*.
- A poset that is both a join semi-lattice and a meet semi-lattice is said to be a *lattice*.
- If  $L$  has a unique top element, we may designate that element by  $\hat{1}$  or  $\hat{1}_L$ .
- If  $L$  has a unique bottom element, we may designate that element by  $\hat{0}$  or  $\hat{0}_L$ .
- If  $L$  is a finite join semi-lattice with a unique bottom element, then  $L$  is a lattice. Similarly, if  $L$  is a finite meet semi-lattice with a unique top element, then  $L$  is a lattice.
- A lattice  $L$  is said to be *bounded* if it has a unique top element  $\hat{1}$  and a unique bottom element  $\hat{0}$ . (These are same element if  $L$  is a singleton.)
- When  $L$  is a bounded lattice, the *proper part* of  $L$  is the poset  $\overline{L} = L \setminus \{\hat{0}, \hat{1}\}$ .
- Suppose  $L$  is a bounded lattice and  $p \in L$ . Then the *complements* of  $p$  are given by the set  $\mathfrak{C}(p) = \{q \in L \mid q \vee p = \hat{1} \text{ and } q \wedge p = \hat{0}\}$ .
- A bounded lattice  $L$  is said to be *noncomplemented* if  $\mathfrak{C}(p) = \emptyset$  for at least one  $p \in L$ . The proper part  $\overline{L}$  of a noncomplemented lattice  $L$  is contractible ([1], Theorem 10.15).
- Suppose  $L$  is a bounded lattice. The elements of  $L$  immediately below  $\hat{1}$  are called *co-atoms*. These are the maximal elements of  $\overline{L}$ . The elements immediately above  $\hat{0}$  are called *atoms*. These are the minimal elements of  $\overline{L}$ .



## A.4 Relations

Let  $R$  be a relation on  $X \times Y$ . We use the following notation and conventions:

- $R$  is a set of pairs, namely a subset of the cross product  $X \times Y$ . It is convenient sometimes to view  $R$  as a matrix of 0s and 1s, perhaps drawn as a matrix of blank and nonblank entries, representing the characteristic function of this set of pairs.
- Even if  $X \neq \emptyset$  and  $Y \neq \emptyset$ , it is possible that  $R = \emptyset$ , in which case we say that  $R$  is an *empty relation*.
- If either  $X = \emptyset$  or  $Y = \emptyset$ , then we say that  $R$  is a *void relation*.

On some occasions, we may treat a void relation  $R$  much like an empty relation, in the sense that we will let the *Dowker complexes* defined below be empty rather than void. That view will sometimes be convenient when  $R$  is derived from some encompassing relation as a link or deletion in a simplicial complex.

- We often refer to elements of  $X$  as *individuals* and elements of  $Y$  as *attributes*.
- $Y_x$  is the set of attributes that individual  $x$  has (in relation  $R$ ). Viewing  $R$  as a matrix, one may think of  $Y_x$  as the row of  $R$  indexed by  $x$ . We say that *the row is blank* when  $Y_x = \emptyset$ .
- $X_y$  is the set of individuals who have attribute  $y$  (in relation  $R$ ). Viewing  $R$  as a matrix, one may think of  $X_y$  as the column of  $R$  indexed by  $y$ . The *column is blank* when  $X_y = \emptyset$ .
- $\Phi_R$  is the Dowker simplicial complex associated with  $R$  whose underlying vertex set is  $Y$ . A nonempty subset  $\gamma$  of  $Y$  is a simplex in  $\Phi_R$  precisely when there exists  $x \in X$  such that  $(x, y) \in R$  for all  $y \in \gamma$ . We refer to  $x$  as a *witness for  $\gamma$* .

When  $R$  is void,  $\Phi_R$  is void as well, except as otherwise indicated in the text.

When  $R$  is nonvoid,  $\Phi_R$  contains the empty simplex. Moreover, we may view  $\Phi_R$  as generated by all the rows of  $R$ . In particular,  $Y_x \in \Phi_R$  for each  $x \in X$ .

- $\Psi_R$  is the Dowker simplicial complex associated with  $R$  whose underlying vertex set is  $X$ . A nonempty subset  $\sigma$  of  $X$  is a simplex in  $\Psi_R$  precisely when there exists  $y \in Y$  such that  $(x, y) \in R$  for all  $x \in \sigma$ . We refer to  $y$  as a *witness for  $\sigma$* .

When  $R$  is void,  $\Psi_R$  is void as well, except as otherwise indicated in the text.

When  $R$  is nonvoid,  $\Psi_R$  contains the empty simplex. Moreover, we may view  $\Psi_R$  as generated by all the columns of  $R$ . In particular,  $X_y \in \Psi_R$  for each  $y \in Y$ .

- There exist homotopy equivalences  $\phi_R : \Psi_R \rightarrow \Phi_R$  and  $\psi_R : \Phi_R \rightarrow \Psi_R$ .

Viewed as poset maps  $\phi_R : \mathfrak{F}(\Psi_R) \rightarrow \mathfrak{F}(\Phi_R)$  and  $\psi_R : \mathfrak{F}(\Phi_R) \rightarrow \mathfrak{F}(\Psi_R)$ , one obtains explicit formulas, sending nonempty simplices to nonempty simplices:

$$\phi_R(\sigma) = \bigcap_{x \in \sigma} Y_x \quad \text{and} \quad \psi_R(\gamma) = \bigcap_{y \in \gamma} X_y.$$

Suppose  $X \neq \emptyset$  and  $Y \neq \emptyset$ . Then the intersections appearing in the previous formulas comprise the witnesses for the respective simplex arguments. Consequently, one may use the formulas more generally as tests for membership in the Dowker complexes:

- For any  $\sigma \subseteq X$ ,  $\sigma \in \Psi_R$  if and only if  $\phi_R(\sigma) \neq \emptyset$ .
- For any  $\gamma \subseteq Y$ ,  $\gamma \in \Phi_R$  if and only if  $\psi_R(\gamma) \neq \emptyset$ .

These tests also make sense for the empty set, that is, when  $\sigma = \emptyset$  or  $\gamma = \emptyset$ . In particular,  $\phi_R(\emptyset) = Y$  and  $\psi_R(\emptyset) = X$ .

- Composing  $\phi_R$  and  $\psi_R$  as  $\psi_R \circ \phi_R : \mathfrak{F}(\Psi_R) \rightarrow \mathfrak{F}(\Psi_R)$  and  $\phi_R \circ \psi_R : \mathfrak{F}(\Phi_R) \rightarrow \mathfrak{F}(\Phi_R)$  produces closure operators. See Appendix B for further details.
- $P_R$  is the *doubly-labeled poset* associated with  $R$  as per Definition 3 on page 20. Each element in  $P_R$  is of the form  $(\sigma, \gamma)$ , with  $\sigma \neq \emptyset$  and  $\gamma \neq \emptyset$ , such that  $\sigma = \psi_R(\gamma)$  and  $\gamma = \phi_R(\sigma)$ .

One may view  $P_R$  either as the image  $(\psi_R \circ \phi_R)(\mathfrak{F}(\Psi_R))$  or as the image  $(\phi_R \circ \psi_R)(\mathfrak{F}(\Phi_R))$ .

- $P_R^+$  is the *Galois lattice* formed from  $P_R$  as per Definition 12 on page 38.
- We sometimes view  $P_R$  as “almost a join-based lattice”, as per Definition 24 on page 47. That amounts to adjoining a single new element  $\hat{1}$  above  $P_R$ , then inducing a join operation on  $P_R \cup \{\hat{1}\}$  from the join operation on  $P_R^+$ . Thus  $P_R \cup \{\hat{1}\}$  is a join semi-lattice. If we further adjoin a new bottom element  $\hat{0}$ , then  $P_R \cup \{\hat{0}, \hat{1}\}$  is a lattice.
- One may speak of the *topology of a relation*: One says that a relation  $R$  has a topological property when any and all of  $\Phi_R$ ,  $\Psi_R$ , and  $\Delta(P_R)$  have that property. (This convention makes sense by Dowker’s Theorem on page 17 and the nature of  $P_R$ .)

## B Basic Tools

This appendix reviews some basic facts about relations, their Dowker complexes, and the Galois connection. Recall the formulas from page 85.

Although we do not always say so explicitly, there are dual statements for the lemmas and corollaries in this appendix, for each of the two perspectives offered by Dowker's Theorem, by inverting the roles of individuals and attributes.

**Lemma 41.** *Let  $R$  be a relation on  $X \times Y$ . Then  $\phi_R$  is inclusion-reversing.*

*Proof.* Let  $\sigma' \subseteq \sigma \subseteq X$ . Then: 
$$\phi_R(\sigma') = \bigcap_{x \in \sigma'} Y_x \supseteq \bigcap_{x \in \sigma} Y_x = \phi_R(\sigma).$$

Just to be careful: if  $\sigma' = \emptyset$ , then  $\phi_R(\sigma') = Y$ , which does indeed contain  $\phi_R(\sigma)$ .  $\square$

Each of  $\phi_R$  and  $\psi_R$  is inclusion-reversing, so  $\phi_R \circ \psi_R$  is inclusion-preserving. Lemmas 42 and 44 establish that  $\phi_R \circ \psi_R$  is a closure operator when viewed as a poset map  $\mathfrak{F}(\Phi_R) \rightarrow \mathfrak{F}(\Phi_R)$ :

**Lemma 42.** *Let  $R$  be a relation on  $X \times Y$ . For all  $\gamma \subseteq Y$ ,  $\gamma \subseteq (\phi_R \circ \psi_R)(\gamma)$ .*

*Proof.*

$$(\phi_R \circ \psi_R)(\gamma) = \bigcap_{x \in \sigma} Y_x, \quad \text{with } \sigma = \bigcap_{y \in \gamma} X_y.$$

The assertion is clear if  $\gamma = \emptyset$  or  $\sigma = \emptyset$ . Otherwise, let  $y \in \gamma$  and  $x \in \sigma$ . Then  $x \in X_y$ , so  $y \in Y_x$ . Since  $x$  is arbitrary in  $\sigma$ , we see that  $y \in (\phi_R \circ \psi_R)(\gamma)$  and thus  $\gamma \subseteq (\phi_R \circ \psi_R)(\gamma)$ .  $\square$

**Corollary 43.** *Let  $R$  be a relation on  $X \times Y$ .*

*If  $\gamma$  is a maximal simplex of  $\Phi_R$ , then  $(\phi_R \circ \psi_R)(\gamma) = \gamma$ .*

*Proof.* When  $\gamma \neq \emptyset$ , this assertion follows from Lemma 42 and maximality of  $\gamma$ . Otherwise, apparently  $\Phi_R = \{\emptyset\}$  and so  $(\phi_R \circ \psi_R)(\emptyset) = \phi_R(X) = \emptyset$  (since  $\phi_R$  must map  $X$  into  $\Phi_R$ ).  $\square$

**Lemma 44.** *Let  $R$  be a relation on  $X \times Y$ .*

*For all  $\gamma \subseteq Y$ ,  $((\phi_R \circ \psi_R) \circ (\phi_R \circ \psi_R))(\gamma) = (\phi_R \circ \psi_R)(\gamma)$ .*

*Proof.* Consider:  $\gamma \xrightarrow{\psi_R} \sigma \xrightarrow{\phi_R} \gamma' \xrightarrow{\psi_R} \sigma' \xrightarrow{\phi_R} \gamma''.$

We need to show that  $\gamma' = \gamma''$ .

By Lemma 42 and its dualization,  $\gamma \subseteq \gamma' \subseteq \gamma''$  and  $\sigma \subseteq \sigma'$ .

By Lemma 41,  $\phi_R$  is inclusion-reversing, so  $\sigma \subseteq \sigma'$  implies  $\gamma' \supseteq \gamma''$ , and thus  $\gamma' = \gamma''$ .

Comment: By the dual of Lemma 41,  $\psi_R$  is inclusion-reversing, so in fact also  $\sigma = \sigma'$ .  $\square$

**Corollary 45.** *Let  $R$  be a relation on  $X \times Y$ . For all  $\sigma \subseteq X$ ,  $(\phi_R \circ \psi_R)(\phi_R(\sigma)) = \phi_R(\sigma)$ .*

*Proof.* This follows from a dual version of the comment at the end of the proof of Lemma 44.  $\square$

**Corollary 46.** *Let  $R$  be a relation on  $X \times Y$ . For all  $x \in X$ ,  $(\phi_R \circ \psi_R)(Y_x) = Y_x$ .*

*Proof.* The assertion follows from Corollary 45, with  $\sigma = \{x\}$ .

(This includes the case  $Y_x = \emptyset$ .)  $\square$

**Lemma 47.** *Let  $R$  be a relation on  $X \times Y$  and suppose  $\eta \subseteq Y$ . The following two conditions are equivalent:*

- (a)  $(\phi_R \circ \psi_R)(\chi) = \chi$ , for every proper subset  $\chi$  of  $\eta$ ,
- (b)  $(\phi_R \circ \psi_R)(\gamma) = \gamma$ , for all  $\gamma$  of the form  $\gamma = \eta \setminus \{y\}$  with  $y \in \eta$ .

*Proof.* Certainly (a) implies (b). Suppose (b) holds, but there is some  $\chi \subsetneq \eta$  such that  $\chi \subsetneq (\phi_R \circ \psi_R)(\chi)$ . Let  $y \in (\phi_R \circ \psi_R)(\chi) \setminus \chi$  and consider  $\gamma = \eta \setminus \{y\}$ .

Observe that  $\chi \subseteq \gamma$ , so  $y \in (\phi_R \circ \psi_R)(\chi) \subseteq (\phi_R \circ \psi_R)(\gamma)$ . Consequently,

$$\eta = \gamma \cup \{y\} \subseteq (\phi_R \circ \psi_R)(\gamma) = \gamma \subsetneq \eta, \quad \text{which is a contradiction.} \quad \square$$

**Definition 48** (Connected). *A relation  $R$  on  $X \times Y$  is connected if  $R$  is connected when viewed as an undirected bipartite graph on the vertex sets  $X$  and  $Y$ .*

**Definition 49** (Tight). *A relation  $R$  on  $X \times Y$  is tight if it has no blank rows or columns.*

**Lemma 50** (Connectedness). *Let  $R$  be a tight relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. Then the following three conditions are equivalent:*

- (a)  $R$  is connected.
- (b)  $\Psi_R$  is path-connected.
- (c)  $\Phi_R$  is path-connected.

*Proof.* We will show that (a) and (b) are equivalent. The proof for (a) and (c) is similar, or one can simply invoke Dowker duality.

I. Suppose  $R$  is connected. Consider two vertices  $x_0$  and  $x_f$  of  $\Psi_R$ . Since  $R$  is connected as a bipartite graph, there exists a path  $x_0, y_1, x_1, y_2, \dots, y_n, x_n = x_f$ . Observe that each  $y_i$  is a witness for the simplex  $\{x_{i-1}, x_i\} \in \Psi_R$ , so in  $\Psi_R$  there exist edges  $\{x_0, x_1\}, \dots, \{x_{n-1}, x_n\}$ . Since  $\Psi_R$  is a simplicial complex, we see that it is path-connected.

II. Suppose  $\Psi_R$  is path-connected. Since  $R$  is tight, each  $y \in Y$  appears as the vertex of an edge  $(x, y)$  in the bipartite graph  $R$ . To show that  $R$  is connected, it therefore is enough to show that any two elements  $x_0$  and  $x_f$  of  $X$  may be connected by a path in the bipartite graph. Since  $R$  is tight,  $x_0$  and  $x_f$  are each vertices of  $\Psi_R$ . Since  $\Psi_R$  is path-connected, there exists a path between  $x_0$  and  $x_f$  in  $\Psi_R$ . Since  $\Psi_R$  is a finite simplicial complex, we can deform that path so that it consists of finitely many edges  $\{x_0, x_1\}, \dots, \{x_{n-1}, x_n\}$ , with each  $x_i$  a vertex of  $\Psi_R$  and  $x_n = x_f$ . Each edge  $\{x_{i-1}, x_i\}$  has some witness  $y_i \in Y$ . So  $x_0, y_1, x_1, y_2, \dots, y_n, x_f$  is a path connecting  $x_0$  and  $x_f$  in the bipartite graph  $R$ .  $\square$

**Lemma 51** (Components). *Let  $R$  be a tight relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. Suppose  $R = R_1 \cup \dots \cup R_\ell$ , with the  $\{R_i\}$  pairwise disjoint and each  $R_i$  a connected component of  $R$  viewed as a bipartite graph on  $X$  and  $Y$ . Then  $X$ ,  $Y$ ,  $\Psi_R$ , and  $\Phi_R$  decompose as follows:*

- (a)  $X = X_1 \cup \dots \cup X_\ell$ , with the  $\{X_i\}$  pairwise disjoint and each  $X_i$  not empty.
- (b)  $Y = Y_1 \cup \dots \cup Y_\ell$ , with the  $\{Y_i\}$  pairwise disjoint and each  $Y_i$  not empty.
- (c)  $R_i$  is the restriction of  $R$  to  $X_i \times Y_i$ , and is tight, for  $i = 1, \dots, \ell$ .
- (d)  $\Psi_R = \Psi_{R_1} \cup \dots \cup \Psi_{R_\ell}$ , with the  $\{\Psi_{R_i}\}$  pairwise disjoint and each  $\Psi_{R_i}$  path-connected.
- (e)  $\Phi_R = \Phi_{R_1} \cup \dots \cup \Phi_{R_\ell}$ , with the  $\{\Phi_{R_i}\}$  pairwise disjoint and each  $\Phi_{R_i}$  path-connected.

*Proof.* Let  $X_i = \{x \mid (x, y) \in R_i \text{ for some } y \in Y\}$  and  $Y_i = \{y \mid (x, y) \in R_i \text{ for some } x \in X\}$ , for  $i = 1, \dots, \ell$ . These sets are nonempty since the components of  $R$  are necessarily nonempty.

To see that  $X_i \cap X_j = \emptyset$  unless  $i = j$ , suppose  $x \in X_i \cap X_j$ . Then  $(x, y) \in R_i$  for some  $y \in Y$  and  $(x, y') \in R_j$  for some  $y' \in Y$ . Since  $R_i$  and  $R_j$  are connected components of  $R$ ,  $i = j$ . Next observe that each  $x$  of  $X$  must appear in some  $X_i$  since  $R$  has no blank rows. Point (a) follows. Point (b) is similar.

For (c), observe that if  $(x, y) \in R_i \subseteq R$  then  $x \in X_i$  and  $y \in Y_i$ , so  $(x, y)$  is in the restriction of  $R$  to  $X_i \times Y_i$ . Conversely, if  $(x, y) \in R$  with  $x \in X_i$  and  $y \in Y_i$ , then  $(x, y) \in R_j$  for some  $j$ . By the previous reasoning,  $i = j$ . Tightness follows by definition of  $X_i$  and  $Y_i$ .

For (d),  $\Psi_{R_i} \subseteq \Psi_R$  since  $R_i \subseteq R$ , for each  $i = 1, \dots, \ell$ . Now suppose  $\emptyset \neq \sigma \in \Psi_R$ . Then there exists  $y \in Y$  such that  $(x, y) \in R$  for every  $x \in \sigma$ . For some  $i$ ,  $y \in Y_i$ . Since  $R_i$  is a connected component of  $R$ ,  $(x, y) \in R_i$  for every  $x \in \sigma$ , so  $\sigma \in \Psi_{R_i}$ . The  $\{\Psi_{R_i}\}$  are pairwise disjoint since the underlying vertex sets  $\{X_i\}$  are pairwise disjoint. Path-connectedness follows from Lemma 50, since each  $R_i$  is tight and connected. Point (e) is similar.  $\square$

**Corollary 52** (Component Maps). *Assume the hypotheses and constructions as in Lemma 51 and its proof. Then:*

$$\begin{aligned} \psi_{R_i}(\gamma) &= \psi_R(\gamma), \quad \text{for each } \emptyset \neq \gamma \in \Phi_{R_i}, \\ \phi_{R_i}(\sigma) &= \phi_R(\sigma), \quad \text{for each } \emptyset \neq \sigma \in \Psi_{R_i}, \quad i = 1, \dots, \ell. \end{aligned}$$

*Proof.* By direct computation:  $\psi_{R_i}(\gamma) = \bigcap_{y \in \gamma} (X_y \cap X_i) = \bigcap_{y \in \gamma} X_y = \psi_R(\gamma)$ .

The second equality comes from the fact that each  $X_y$  can touch only  $X_i$ , since  $R_i$  is a connected component of  $R$ . The argument for the  $\phi_{\dots}$  maps is similar.  $\square$

**Corollary 53** (Component Privacy). *Assume the hypotheses and constructions as in Lemma 51 and its proof. Let  $i \in \{1, \dots, \ell\}$ .*

*If  $\psi_R \circ \phi_R$  is the identity on  $\Psi_R$  and  $Y_i \not\subseteq \Phi_{R_i}$ , then  $\psi_{R_i} \circ \phi_{R_i}$  is the identity on  $\Psi_{R_i}$ .*

*If  $\phi_R \circ \psi_R$  is the identity on  $\Phi_R$  and  $X_i \not\subseteq \Psi_{R_i}$ , then  $\phi_{R_i} \circ \psi_{R_i}$  is the identity on  $\Phi_{R_i}$ .*

*Proof.* Suppose  $\emptyset \neq \sigma \in \Psi_{R_i}$ , then  $\emptyset \neq \phi_{R_i}(\sigma) \in \Phi_{R_i}$ , so by Corollary 52,  $(\psi_{R_i} \circ \phi_{R_i})(\sigma) = (\psi_R \circ \phi_R)(\sigma) = \sigma$ . And  $(\psi_{R_i} \circ \phi_{R_i})(\emptyset) = \psi_{R_i}(Y_i) = \emptyset$ , since  $Y_i \not\subseteq \Phi_{R_i}$ .

The argument for  $\phi_{R_i} \circ \psi_{R_i}$  is similar.  $\square$

## C Links and Inference

This appendix provides some technical tools for modeling inference, particularly in links, ending with some instances in which inference is unavoidable.

**Intuition:** The link  $\text{Lk}(\Phi_R, \gamma)$  of a set of attributes  $\gamma$  in the Dowker complex  $\Phi_R$  can be understood as a description of what may yet be observed or inferred, *conditional* on having already observed  $\gamma$ .

**Lemma 54.** *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. Suppose  $\gamma \in \Phi_R$ . Define relation  $Q$  as a restriction of  $R$  by*

$$Q = R|_{\sigma \times \bar{Y}}, \quad \text{with } \sigma = \psi_R(\gamma) \quad \text{and} \quad \bar{Y} = \bigcup_{x \in \sigma} Y_x \setminus \gamma.$$

*Then  $\text{Lk}(\Phi_R, \gamma) = \Phi_Q$ , as collections of simplices (i.e., ignoring underlying vertex sets).*

*(Observe that  $\sigma \neq \emptyset$ . If  $\bar{Y} = \emptyset$ , then technically  $Q$  is void, but it is convenient to let both  $\Phi_Q$  and  $\Psi_Q$  be instances of the empty complex  $\{\emptyset\}$ . — In a standard link, one might define  $\bar{Y} = Y \setminus \gamma$ . With  $\bar{Y}$  as above,  $Q$  always discards blank columns of  $R$ , even when  $\gamma = \emptyset$ .)*

*Proof.* Observe that  $\gamma \subseteq Y_x$  if and only if  $x \in \sigma$ .

We discuss the case  $\bar{Y} = \emptyset$  separately, for clarity. We need to show that  $\text{Lk}(\Phi_R, \gamma) = \{\emptyset\}$ . If  $\text{Lk}(\Phi_R, \gamma) \neq \{\emptyset\}$ , then there exists some  $\bar{y} \in \text{verts}(\text{Lk}(\Phi_R, \gamma))$ . By definition of link,  $\bar{y} \notin \gamma$  and there exists  $\bar{x} \in X$  such that  $(\bar{x}, y) \in R$  for all  $y \in \gamma \cup \{\bar{y}\}$ . That means  $\bar{x} \in \sigma$ , so  $\bar{y} \in \bar{Y}$ , a contradiction.

The converse is true as well: If  $\text{Lk}(\Phi_R, \gamma) = \{\emptyset\}$ , then  $\bar{Y} = \emptyset$ . For if some  $x \in \sigma$  has an attribute  $y$  in addition to all those in  $\gamma$ , then  $y$  would be a vertex in the link.

Now suppose  $\bar{Y} \neq \emptyset$ :

I. If  $\xi \in \text{Lk}(\Phi_R, \gamma)$ , then  $\xi \cap \gamma = \emptyset$  and there exists  $x \in X$  such that  $(x, y) \in R$  for every  $y \in \xi \cup \gamma$ . So  $\xi \subseteq Y_x \setminus \gamma$  and  $x \in \psi_R(\gamma) = \sigma$ . Thus  $(x, y) \in Q$  for every  $y \in \xi$ , meaning  $\xi \in \Phi_Q$ .

II. Conversely, if  $\xi \in \Phi_Q$ , then there exists  $x \in \sigma$  such that  $(x, y) \in Q \subseteq R$  for every  $y \in \xi$ . By definition of  $\sigma$ ,  $(x, y) \in R$  for every  $y \in \gamma$ . Combining these two assertions, we see that  $(x, y) \in R$  for every  $y \in \xi \cup \gamma$ . And  $\xi \cap \gamma = \emptyset$  since  $\xi \subseteq \bar{Y}$ . So  $\xi \in \text{Lk}(\Phi_R, \gamma)$ .  $\square$

**Comment:** There is a dual version of this lemma for links of individuals  $\sigma$ , modeling  $\text{Lk}(\Psi_R, \sigma)$  as  $\Psi_Q$  for an appropriate relation  $Q$ . We see an instance of that in Theorem 9 on page 100, with  $\sigma$  consisting of a single individual  $x$ .

With notation and construction as in Lemma 54, the following formulas hold, assuming  $\bar{Y} \neq \emptyset$ :

- Suppose  $\xi \subseteq \bar{Y}$  and define  $\tau = \xi \cup \gamma$ . Then

$$\psi_Q(\xi) = \bigcap_{y \in \xi} (X_y \cap \sigma) = \left( \bigcap_{y \in \xi} X_y \right) \cap \left( \bigcap_{y \in \gamma} X_y \right) = \bigcap_{y \in (\xi \cup \gamma)} X_y = \psi_R(\tau).$$

Notes: We allow  $\xi = \emptyset$ , since  $\psi_Q(\emptyset) = \sigma = \psi_R(\gamma)$ . We do not require  $\xi \in \Phi_Q$ . The equalities hold regardless. Of course,  $\xi \in \Phi_Q$  if and only if  $\psi_Q(\xi) \neq \emptyset$ .

- Suppose  $\emptyset \neq \kappa \subseteq \sigma$ . Then

$$\phi_Q(\kappa) = \bigcap_{x \in \kappa} (Y_x \cap \bar{Y}) = \left( \bigcap_{x \in \kappa} Y_x \right) \setminus \gamma = \phi_R(\kappa) \setminus \gamma.$$

And thus also  $\phi_R(\kappa) = \phi_Q(\kappa) \cup \gamma$ , since  $\gamma \subseteq Y_x$  for all  $x \in \sigma$ .

Notes: Here we do *not* allow  $\kappa = \emptyset$ , since  $\phi_Q(\emptyset) = \bar{Y}$  whereas  $\phi_R(\emptyset) = Y$ . It need not be true that  $Y = \bar{Y} \cup \gamma$ . Again,  $\kappa \in \Psi_Q$  if and only if  $\phi_Q(\kappa) \neq \emptyset$ , this valid also for  $\kappa = \emptyset$ .

Comment: If  $\bar{Y} = \emptyset$ , the previous formulas still hold, albeit trivially. However, testing for membership in  $\Psi_Q$  via the question “Is  $\phi_Q(\kappa)$  nonempty?” no longer makes sense.

**Lemma 55.** *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. Suppose  $\gamma \subseteq Y$ . Then  $\text{dl}(\Phi_R, \gamma) = \Phi_{Q'}$ , with  $Q'$  formed from  $R$  by removing the columns corresponding to  $\gamma$ , that is,  $Q' = R|_{X \times (Y \setminus \gamma)}$ . (Here we let  $\Psi_{Q'}$  and  $\Phi_{Q'}$  each be the empty complex if  $\gamma = Y$ .)*

*Proof.* An individual  $x \in X$  is a witness to a set of attributes  $\xi \subseteq Y \setminus \gamma$  in  $R$  if and only if  $x$  is a witness to  $\xi$  in  $Q'$ .  $\square$

With notation and construction as in Lemma 55, the following formulas hold, assuming  $\gamma \neq Y$ :

- If  $\xi \subseteq (Y \setminus \gamma)$ , then  $\psi_{Q'}(\xi) = \bigcap_{y \in \xi} X_y = \psi_R(\xi)$ .
- If  $\kappa \subseteq X$ , then  $\phi_{Q'}(\kappa) = \bigcap_{x \in \kappa} (Y_x \setminus \gamma) = \phi_R(\kappa) \setminus \gamma$ .

Caution: It need *not* be true that  $\phi_R(\kappa) = \phi_{Q'}(\kappa) \cup \gamma$ .

Comments: (1) The first formula holds for  $\xi = \emptyset$  and the second formula holds for  $\kappa = \emptyset$ . (2) The simplex tests hold:  $\xi \in \Phi_{Q'}$  if and only if  $\psi_{Q'}(\xi) \neq \emptyset$ ; and  $\kappa \in \Psi_{Q'}$  if and only if  $\phi_{Q'}(\kappa) \neq \emptyset$ . (3) If  $\gamma = Y$ , the formulas still hold, but testing for membership in  $\Psi_{Q'}$  via the question “Is  $\phi_{Q'}(\kappa)$  nonempty?” no longer makes sense.

**Recall:** A relation  $R$  preserves attribute privacy when the closure operator  $\phi_R \circ \psi_R$  is the identity on  $\Phi_R$  and it preserves association privacy when the closure operator  $\psi_R \circ \phi_R$  is the identity on  $\Psi_R$ .

**Lemma 56.** *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. Suppose  $\gamma \in \Phi_R$ .*

*If  $\phi_R \circ \psi_R$  is the identity on  $\Phi_R$ , then the corresponding closure operators for the relations modeling  $\text{Lk}(\Phi_R, \gamma)$  and  $\text{dl}(\Phi_R, \gamma)$  are also identities.*

Technicality: The operators are formally defined as self-maps on the face posets of the simplicial complexes mentioned in the lemma, but we can extend each operator to the empty simplex and therefore think of it as a self-map on a simplicial complex viewed as a collection of simplices.

*Proof.* Define  $Q$  as in Lemma 54. That lemma tells us  $\Phi_Q = \text{Lk}(\Phi_R, \gamma)$ .

Given  $\xi \in \Phi_Q$ , let  $\tau = \xi \cup \gamma$  and calculate:

$$(\phi_Q \circ \psi_Q)(\xi) = \phi_Q(\psi_R(\tau)) = \phi_R(\psi_R(\tau)) \setminus \gamma = \tau \setminus \gamma = \xi.$$

Define  $Q'$  as in Lemma 55. That lemma tells us  $\Phi_{Q'} = \text{dl}(\Phi_R, \gamma)$ .

Given  $\xi \in \Phi_{Q'}$ , calculate:

$$(\phi_{Q'} \circ \psi_{Q'})(\xi) = \phi_{Q'}(\psi_R(\xi)) = \phi_R(\psi_R(\xi)) \setminus \gamma = \xi \setminus \gamma = \xi. \quad \square$$

Here is a variation, in which one again computes a link of attributes but then considers the closure operator on the dual complex, i.e., within the space of individuals:

**Lemma 57.** *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. Suppose  $\gamma \in \Phi_R$ .*

*Let  $Q$ ,  $\sigma$ , and  $\bar{Y}$  be as in the construction of Lemma 54. Assume  $|\sigma| > 1$  and  $\bar{Y} \neq \emptyset$ .*

*If  $\psi_R \circ \phi_R$  is the identity on  $\Psi_R$ , then  $\psi_Q \circ \phi_Q$  is the identity on  $\Psi_Q$ .*

*Proof.* Suppose  $\emptyset \neq \kappa \in \Psi_Q$ . Observe that  $\gamma \subseteq \phi_R(\kappa)$  and calculate:

$$(\psi_Q \circ \phi_Q)(\kappa) = \psi_Q(\phi_R(\kappa) \setminus \gamma) = \psi_R(\phi_R(\kappa)) = \kappa.$$

Additionally,

$$\begin{aligned} (\psi_Q \circ \phi_Q)(\emptyset) &= \psi_Q(\bar{Y}) = \psi_R(\bar{Y} \cup \gamma) = \psi_R\left(\bigcup_{x \in \sigma} Y_x\right) = \\ &= \bigcap_{x \in \sigma} \psi_R(Y_x) = \bigcap_{x \in \sigma} (\psi_R \circ \phi_R)(\{x\}) = \bigcap_{x \in \sigma} \{x\} = \emptyset. \end{aligned}$$

The last equality holds since  $|\sigma| > 1$ . In short,  $(\psi_Q \circ \phi_Q)(\kappa) = \kappa$  for all  $\kappa \in \Psi_Q$ .  $\square$

**Comment:** When  $\bar{Y} = \emptyset$ , we take  $\Psi_Q$  and  $\Phi_Q$  to be the empty simplicial complex  $\{\emptyset\}$ . It is sensible to say that  $\phi_Q \circ \psi_Q$  is the identity on  $\Phi_Q$  since  $\phi_Q(\psi_Q(\emptyset)) = \phi_Q(\sigma) = \emptyset$ . It could be confusing to say that  $\psi_Q \circ \phi_Q$  is the identity on  $\Psi_Q$  since  $\psi_Q(\phi_Q(\emptyset)) = \psi_Q(\bar{Y}) = \psi_Q(\emptyset) = \sigma$ , though perhaps one could argue that there should be no association inference in  $Q$  since there are no attributes.

**Corollary 58.** *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. Suppose  $\gamma \in \Phi_R$ .*

*Let  $Q$  and  $\bar{Y}$  be as in the construction of Lemma 54. Assume  $\bar{Y} \neq \emptyset$ .*

*If  $R$  preserves both attribute and association privacy, then so does  $Q$ .*

*Proof.* Relation  $Q$  preserves attribute privacy by Lemma 56. Let  $\sigma = \psi_R(\gamma)$ . If we can show that  $|\sigma| > 1$ , then  $Q$  preserves association privacy by Lemma 57.

Observe that  $|\sigma| > 0$ , since  $\gamma \in \Phi_R$ . If  $\psi(\gamma)$  consists of a single individual  $x \in X$ , then

$$\gamma = (\phi_R \circ \psi_R)(\gamma) = \phi_R(\sigma) = Y_x = \bar{Y} \cup \gamma,$$

which is impossible for nonempty  $\bar{Y}$ .  $\square$



The following lemma formalizes the intuition that a set of attributes  $\gamma$  implies another attribute  $y$  precisely when the columns corresponding to  $\gamma$  have nonempty intersection and that intersection is a subset of the column corresponding to  $y$ .

**Lemma 59.** *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty.*

*$R$  preserves attribute privacy if and only if the following condition is true:*

*For all  $\gamma \in \Phi_R$  and all  $y \in Y$ , if  $\psi_R(\gamma) \subseteq \psi_R(\{y\})$  then  $y \in \gamma$ .*

*Proof.* I. Suppose there exist  $\gamma \in \Phi_R$  and  $y \in Y$  such that  $\psi_R(\gamma) \subseteq \psi_R(\{y\})$  but  $y \notin \gamma$ . Since  $\phi_R \circ \psi_R$  is a closure operator,  $y \in (\phi_R \circ \psi_R)(\{y\})$  and  $\gamma \subseteq (\phi_R \circ \psi_R)(\gamma)$ . Now observe that  $(\phi_R \circ \psi_R)(\{y\}) \subseteq (\phi_R \circ \psi_R)(\gamma)$  by supposition and because  $\phi_R$  is inclusion-reversing. Consequently,  $(\phi_R \circ \psi_R)(\gamma)$  must be a proper superset of  $\gamma$ , telling us there is attribute inference.

II. If there is attribute inference, then for some  $\gamma \in \Phi_R$ ,  $\gamma \subsetneq (\phi_R \circ \psi_R)(\gamma)$ . Pick some  $y \in (\phi_R \circ \psi_R)(\gamma) \setminus \gamma$ . Then  $y \notin \gamma$  but

$$\psi_R(\gamma) = \psi_R((\phi_R \circ \psi_R)(\gamma)) \subseteq \psi_R((\phi_R \circ \psi_R)(\gamma) \setminus \gamma) \subseteq \psi_R(\{y\}).$$

(The equality holds by associativity of  $\circ$  and the dual version of Corollary 45 on page 87. The two subset relations hold by inclusion-reversal of  $\psi_R$ .)

(Technical comment: In both parts above,  $\gamma = \emptyset$  is permissible.) □

Recall the following definition:

**Definition 6** (Unique Identifiability). *Let  $R$  be a relation on  $X \times Y$  and suppose  $x \in X$ . We say that  $x$  is uniquely identifiable via relation  $R$  when  $\psi_R(Y_x) = \{x\}$ .*

Comment: It is entirely possible that one or more proper subsets  $\gamma$  of  $Y_x$  already identifies  $x$ , meaning  $\psi_R(\gamma) = \{x\}$ . Certainly  $x$  is uniquely identifiable in that case. Moreover, the attributes  $Y_x \setminus \gamma$  can be inferred from  $\gamma$ .

**Lemma 60.** *Let  $R$  be a relation on  $X \times Y$  that preserves attribute privacy. Let  $x \in X$ . Then no proper subset of  $Y_x$  identifies  $x$ .*

*Proof.* Suppose for some  $x \in X$  and some  $\gamma \subsetneq Y_x$ ,  $\psi_R(\gamma) = \{x\}$ . A contradiction ensues:

$$\gamma \subsetneq Y_x = \phi_R(\{x\}) = (\phi_R \circ \psi_R)(\gamma) = \gamma. \quad \square$$

We turn now to proving the assertions of Section 5, regarding free faces.

**Lemma 61.** *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. If  $\Phi_R$  contains no free faces, then  $R$  preserves attribute privacy.*

*Proof.* We will show that  $\phi_R \circ \psi_R$  is the identity on  $\Phi_R$ .

As usual,  $(\phi_R \circ \psi_R)(\emptyset) = \phi_R(X)$ . We therefore need to show that  $\phi_R(X) = \emptyset$ . Observe that every maximal simplex of  $\Phi_R$  contains  $\phi_R(X)$ , since any witness for such a simplex must have all the attributes in  $\phi_R(X)$ . Pick some maximal simplex  $\eta$  of  $\Phi_R$  and consider  $\gamma = \eta \setminus \phi_R(X)$ . Let  $\eta'$  be any maximal simplex of  $\Phi_R$  containing  $\gamma$ . Then

$$\eta = \gamma \cup \phi_R(X) \subseteq \eta' \cup \phi_R(X) = \eta'.$$

So  $\eta = \eta'$  by maximality. Since  $\Phi_R$  has no free faces,  $\gamma$  cannot be a proper subset of  $\eta$ , meaning  $\phi_R(X) = \emptyset$ , as desired.

Now consider  $\emptyset \neq \gamma \in \Phi_R$ . Suppose  $\gamma$  is a proper subset of  $(\phi_R \circ \psi_R)(\gamma)$ . By Corollary 43 and Lemma 47 on pages 87 and 88, respectively, we can assume without loss of generality that  $\gamma = \eta \setminus \{y\}$  for some maximal  $\eta$  of  $\Phi_R$  and some  $y \in \eta$ . Observe that

$$\eta \setminus \{y\} = \gamma \subsetneq (\phi_R \circ \psi_R)(\gamma) \subseteq (\phi_R \circ \psi_R)(\eta) = \eta,$$

so  $\eta = (\phi_R \circ \psi_R)(\gamma)$ . Now let  $\eta'$  be any maximal simplex of  $\Phi_R$  containing  $\gamma$ . Then

$$\eta = (\phi_R \circ \psi_R)(\gamma) \subseteq (\phi_R \circ \psi_R)(\eta') = \eta'.$$

(Note: The last equality in each of the lines of comparisons above follows from Corollary 43 by maximality.)

So  $\eta = \eta'$  by maximality. That says  $\gamma$  is a free face of  $\Phi_R$ , a contradiction.  $\square$

The converse of Lemma 61 need not hold if there exists an individual who can hide, with attributes that form a strict subset of some other individual's attributes. However:

**Lemma 62.** *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. If  $R$  preserves attribute privacy and if every  $x \in X$  is uniquely identifiable via  $R$ , then  $\Phi_R$  contains no free faces.*

*Proof.* Suppose that  $\gamma$  is a free face of  $\Phi_R$ . We can assume without loss of generality that  $\gamma = \eta \setminus \{y\}$  for some maximal  $\eta \in \Phi_R$  and  $y \in \eta$ . Since a Dowker attribute complex is generated by the rows of the underlying relation, it must be that  $\eta = Y_x$  for at least one  $x \in X$ . By Lemma 60, there is at least one  $x'$  besides  $x$  in  $\psi_R(\gamma)$ . Then

$$\gamma = (\phi_R \circ \psi_R)(\gamma) \subseteq \phi_R(\{x, x'\}) = Y_x \cap Y_{x'}.$$

Since we have assumed that  $\gamma$  is free and  $Y_x$  is maximal, we see that  $Y_{x'}$  must be a subset of  $Y_x$ . That means  $x'$  is not uniquely identifiable, a contradiction.

(Technical comment:  $\gamma = \emptyset$  is permissible throughout this argument.)  $\square$

The following lemma will help us later in Appendix E, to establish the remaining assertions of Sections 5 and 7.

**Lemma 63.** *Let  $R$  be a relation on  $X \times Y$  such that  $|X| = |Y| > 1$ . If  $R$  has no blank columns and preserves attribute privacy, then every  $x \in X$  is uniquely identifiable via  $R$ .*

*Proof.* The proof is by induction on  $n = |X| = |Y|$ .

I. The base case  $n = 2$  implies that  $R$  is isomorphic to

$R$	$y_1$	$y_2$
$x_1$	•	
$x_2$		•

(Any other type of  $2 \times 2$  relation without blank columns would allow for attribute inference.)

Each  $x_i$  is uniquely identifiable in  $R$  above.

II. For the induction step, assume that, for some  $n > 2$ , the lemma holds for all relations with  $X$  and  $Y$  spaces of size strictly less than  $n$ . We need to establish the lemma for all relations with  $X$  and  $Y$  spaces of size  $n$ .

**Subclaim:**  $R$  has no blank rows.

To see this, suppose that  $Y_{\tilde{x}} = \emptyset$  for some  $\tilde{x} \in X$ . Let  $Q$  be the restriction of  $R$  to  $X' \times Y$ , with  $X' = X \setminus \{\tilde{x}\}$ . There is no significant difference between  $R$  and  $Q$ ; in particular,  $Q$  also preserves attribute privacy.

(Perhaps the empty simplex is slightly tricky:  $(\phi_Q \circ \psi_Q)(\emptyset) = \bigcap_{x \in X'} Y_x$ . If this intersection is nonempty, it contains some  $y_1 \in Y$ . Pick  $y_2 \in Y$  with  $y_2 \neq y_1$ ; this is possible since  $|Y| > 2$ . Note that  $X_{y_2} \subseteq X'$ , so  $y_1 \in \bigcap_{x \in X'} Y_x \subseteq \bigcap_{x \in X_{y_2}} Y_x = (\phi_R \circ \psi_R)(\{y_2\}) = \{y_2\}$ , a contradiction. So  $(\phi_Q \circ \psi_Q)(\emptyset) = \emptyset$ .)

Now let  $Q'$  be the further restriction of  $R$  to  $X' \times Y'$ , where  $Y' = Y \setminus \tilde{y}$ , with  $\tilde{y}$  any element of  $Y$ . By Lemma 56 on page 91,  $Q'$  preserves attribute privacy. The underlying  $X$  and  $Y$  spaces of  $Q'$  each have size  $n - 1$  and  $Q'$  has no blank columns. The induction hypothesis therefore tells us that every individual in  $X'$  is uniquely identifiable via  $Q'$ . Bearing in mind that  $\tilde{x}$  does not appear in any  $X_y$ , one sees that for each  $x \in X'$ , there is some  $\gamma \subseteq Y'$  such that  $\bigcap_{y \in \gamma} X_y = \{x\}$ . That intersection is a column vector all of whose entries are 0 (blank) except for the entry indexed by  $x$ . Since  $R$  preserves attribute privacy and  $x$  is arbitrary in  $X'$ , Lemma 59 implies that in fact  $X_{\tilde{y}} = \emptyset$ , contradicting the assumption that  $R$  has no blank columns.

Next, pick  $\bar{x} \in X$ . We will show that  $\bar{x}$  is uniquely identifiable via  $R$ . Without loss of generality, write  $R$  as in Figure 51 (“blank” entries are indicated by “0”s):

Specifically, pick some  $\bar{y} \in Y$  such that  $(\bar{x}, \bar{y}) \in R$ . This is possible since  $R$  has no blank rows. Then decompose  $X = X_1 \cup X_2$ , with  $X_1 = X_{\bar{y}}$  and  $X_2 = X \setminus X_1$ . Since  $R$  preserves attribute privacy,  $X_2 \neq \emptyset$ .

	R	$\bar{y}$	$Y_1$	$Y_2$
$X_1$	$\bar{x}$	$\vdots$	Q	<b>0</b>
$X_2$	<b>0</b>		A	B

Figure 51: Relation  $R$  decomposed into blocks for the proof of Lemma 63.

Let  $Q$  model  $\text{Lk}(\Phi_R, \bar{y})$ . So  $Q$  is  $R$  restricted to  $X_1 \times Y_1$ , with  $Y_1 = \bigcup_{x \in X_1} Y_x \setminus \{\bar{y}\}$ . If  $Y_1 \neq \emptyset$ , then  $Q$  preserves attribute privacy, by Lemma 56, and  $Q$  has no blank columns.

Now write  $Y$  as the disjoint union  $Y = \{\bar{y}\} \cup Y_1 \cup Y_2$ , with  $Y_2 = Y \setminus (Y_1 \cup \{\bar{y}\})$ .

Observe that no element of  $X_2$  has attribute  $\bar{y}$ . Observe further that every element in  $X_1$  has attribute  $\bar{y}$  but has no attributes in  $Y_2$ , by construction.

Let  $A$  be the restriction of  $R$  to  $X_2 \times Y_1$  and let  $B$  be the restriction of  $R$  to  $X_2 \times Y_2$ .

If  $Y_2 \neq \emptyset$ , then  $B$  has no blank columns and  $\Phi_B = \text{dl}(\Phi_R, Y_1 \cup \{\bar{y}\})$ . If  $|Y_2| \geq 2$ , then the blank rows indexed by  $X_1$  that remain after deleting from  $R$  the columns indexed by  $Y_1 \cup \{\bar{y}\}$  are irrelevant and so  $B$  preserves attribute privacy (by Lemma 56 and by an argument similar to that appearing in the proof of the Subclaim on page 95).

Let's look at some cases:

- $|Y_2| \geq |X_2| = 1$ : Then any attribute of  $Y_2$  identifies the one element of  $X_2$ . Since  $R$  preserves attribute privacy, this implies both that  $|Y_2| = 1$  and that relation  $A$  is blank. Consequently, every attribute in  $Y_1$  implies  $\bar{y}$  in  $R$ . Since  $R$  preserves attribute privacy, we conclude that  $Y_1 = \emptyset$ . That means we are actually in the base case, with  $n = 2$ .
- $|Y_2| > |X_2| \geq 2$ : By removing some columns of  $B$ , we obtain a square relation to which we can apply the induction hypothesis. That means every  $x \in X_2$  is uniquely identifiable by the remaining columns. Since  $B$  preserves attribute privacy that means the columns removed must have been blank, a contradiction.
- $|Y_2| = |X_2| \geq 2$ : We can apply the induction hypothesis directly to  $B$ . That again tells us that every  $x \in X_2$  is uniquely identifiable by columns of  $Y_2$ , both in  $B$  and in  $R$ . We conclude that relation  $A$  must be blank and so  $Y_1 = \emptyset$ , arguing as above. Thus  $|X_2| = |Y_2| = n - 1$ , implying  $|X_1| = 1$ . So  $\bar{y}$  uniquely identifies  $\bar{x}$ , as desired.
- $|Y_2| < |X_2|$ : This means  $|Y_1| \geq |X_1|$ . Additionally,  $|X_1| \geq 2$ , as otherwise  $\bar{y}$  implies all the attributes of  $Y_1$ . If actually  $|Y_1| > |X_1|$ , then we could argue as above to see that some columns of  $Q$  are blank, contrary to the construction of  $Q$ . So we have that  $|Y_1| = |X_1| \geq 2$  and the induction hypothesis applies. Consequently  $\bar{x}$  is uniquely identifiable via  $Q$ , say as  $\{\bar{x}\} = \psi_Q(\gamma)$ , for some  $\gamma \subseteq Y_1$ . If we adjoin  $\bar{y}$ , we get that  $\psi_R(\gamma \cup \{\bar{y}\}) = \psi_Q(\gamma) = \{\bar{x}\}$ , as desired.  $\square$

**Theorem 64** (Too Many Attributes). *Let  $R$  be a relation on  $X \times Y$  with no blank columns. Suppose  $|Y| > |X| \geq 1$ . Then  $R$  does not preserve attribute privacy.*

*Proof.* The proof is a corollary to Lemma 63:

If  $|Y| > |X| = 1$ , then any one element of  $Y$  implies all the others.

Otherwise, suppose  $R$  preserves attribute privacy. We have  $|Y| > |X| > 1$ , so we can delete some columns of  $R$  and apply Lemma 63 to the resulting relation. Every element of  $X$  is therefore uniquely identifiable via the columns retained. Consequently, either there is attribute inference in  $R$  or the discarded columns were blank, a contradiction.  $\square$

**Comment:** One implication of this result and those in Appendix E is the old detective show mantra “eliminate suspects”: Reduce the number of relevant individuals sufficiently, and some attribute inference is assured. This amounts to moving from relation  $R$  to a subrelation  $Q$  representing  $\text{dl}(\Psi_R, \sigma)$ , with  $\sigma$  a *set* of “eliminated suspects”.

## D Inference Hardness

We have so far spoken mainly of privacy preservation overall in a relation. One can also focus on a single individual:

**Definition 65** (Individual Privacy). *Let  $R$  be a relation on  $X \times Y$  and suppose  $x \in X$ .*

*We say that  $R$  preserves attribute privacy for  $x$  whenever  $(\phi_R \circ \psi_R)(\gamma) = \gamma$  for all  $\gamma \subseteq Y_x$ .*

We have seen the following basic result within the proofs of other lemmas:

**Lemma 66.** *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. Let  $x \in X$ . Then:*

*$R$  preserves attribute privacy for  $x$*

*if and only if*

*$(\phi_R \circ \psi_R)(\gamma) = \gamma$ , for all  $\gamma$  of the form  $\gamma = Y_x \setminus \{y\}$ , with  $y \in Y_x$ .*

*Proof.* I. If  $R$  preserves attribute privacy for  $x$ , then the condition is satisfied by definition.

II. Suppose  $R$  does not preserve attribute privacy for  $x$ . Then for some  $\eta \subseteq Y_x$ ,  $\eta \subsetneq (\phi_R \circ \psi_R)(\eta)$ . We know  $(\phi_R \circ \psi_R)(Y_x) = Y_x$  by Corollary 46 on page 87, so by Lemma 47 on page 88 we can assume that  $\eta = Y_x \setminus \{y\}$ , for some  $y \in Y_x$ .  $\square$

Lemma 66 tells us that it is fairly easy to check whether an individual's attribute privacy is preserved. One merely needs to check whether any one attribute is implied by all the remaining attributes. That may be done quickly since the maps  $\phi_R$  and  $\psi_R$  amount to set intersections. Harder is finding a smallest set of attributes that implies another of the individual's attributes.

Influenced by Lemma 59 on page 93, we formulate the following problem:

**Definition 67** (Minimal Inference). *MININF is the following decision problem:*

*Given relation  $R$  on  $X \times Y$ ,  $x \in X$ ,  $y \in Y$ , and  $k \geq 0$ , is there a simplex  $\gamma \in \Phi_R$  with  $\gamma \subseteq Y_x \setminus \{y\}$  such that  $|\gamma| \leq k$  and  $\psi_R(\gamma) \subseteq \psi_R(\{y\})$ ?*

**Lemma 68.** *MININF is NP-complete*

*Proof.* (A) Observe that the problem lies in NP: Given some  $\gamma$ , one can verify the stated conditions in polynomial time. The verifications amount to set intersection, cardinality, and subset computations, drawn from the columns and one row of  $R$ .

(B) We will establish NP-hardness by a reduction from SET COVER. Recall: Given a collection of sets  $\{S_1, \dots, S_m\}$ , SET COVER asks whether there is some subcollection of size at most  $k$  such that the union of the subcollection is the overall union (often called the *universe*).

Given an instance of the SET COVER problem, we define the following relation:

- $X = \{x_0\} \cup \bigcup_{i=1}^m S_i$ , with  $x_0$  a new element distinct from any elements in the sets  $S_i$ .
- $Y = \{0, 1, \dots, m\}$ .
- $R = (\{x_0\} \times Y) \cup \bigcup_{i=1}^m \{(x, i) \in X \times Y \mid x \in X \setminus S_i\}$ .

In words: The 0<sup>th</sup> column of  $R$  is the singleton set  $\{x_0\}$  and the  $i^{\text{th}}$  column of  $R$ , for  $i = 1, \dots, m$ , is  $X \setminus S_i$ , i.e., the complement of  $S_i$  in the original set cover universe, but now with  $x_0$  added. The row for  $x_0$  has entries for all possible attributes. All other rows have no entry in column 0.

**Reduction:** Given an instance of SET COVER, we transform it into an instance of MININF using the relation  $R$  given above and by letting  $x = x_0$  and  $y = 0$ . The parameter  $k$  is the same for both problems. Observe that  $Y_x \setminus \{y\} = \{1, \dots, m\}$ .

Observe further that  $|X| = |\bigcup_{i=1}^m S_i| + 1 = n + 1$ ,  $|Y| = m + 1$ , with  $n$  the number of elements in the set cover universe and  $m$  the number of subsets specified for the set cover problem. The reduction can therefore be computed in polynomial time.

To complete the proof, we will establish the following:

**Claim:** The answer to SET COVER is “yes” if and only if the answer to MININF is “yes”.

I. A “yes” answer to SET COVER means that there is some set of indices  $\gamma \subseteq \{1, \dots, m\}$ , with  $|\gamma| \leq k$  such that  $\bigcup_{j \in \gamma} S_j = \bigcup_{i=1}^m S_i$ . Therefore, since  $0 \notin \gamma$ ,

$$\psi_R(\gamma) = \bigcap_{j \in \gamma} X_j = \bigcap_{j \in \gamma} (X \setminus S_j) = X \setminus \left( \bigcup_{j \in \gamma} S_j \right) = X \setminus \left( \bigcup_{i=1}^m S_i \right) = \{x_0\} = \psi_R(\{0\}) = \psi_R(\{y\}).$$

In other words,  $\emptyset \neq \psi_R(\gamma) \subseteq \psi_R(\{y\})$  with  $\gamma \subseteq Y_x \setminus \{y\}$  and  $|\gamma| \leq k$ , meaning that the answer to MININF is “yes” as well.

II. A “yes” answer to MININF means there is some  $\gamma \subseteq \{1, \dots, m\}$  such that  $|\gamma| \leq k$  and  $\emptyset \neq \psi_R(\gamma) \subseteq \psi_R(\{y\})$ . Observe that  $\psi_R(\{y\}) = \psi_R(\{0\}) = \{x_0\}$  and that

$$\psi_R(\gamma) = \bigcap_{j \in \gamma} X_j = \bigcap_{j \in \gamma} (X \setminus S_j) = X \setminus \left( \bigcup_{j \in \gamma} S_j \right).$$

The middle equality holds as before because  $0 \notin \gamma$ .

So we see that  $x_0 \in X \setminus \left( \bigcup_{j \in \gamma} S_j \right) \subseteq \{x_0\}$ , telling us

$$\bigcup_{j \in \gamma} S_j = X \setminus \{x_0\} = \bigcup_{i=1}^m S_i.$$

That means  $\gamma$  describes a set of indices sought for by SET COVER, with  $|\gamma| \leq k$ , so the answer to SET COVER is also “yes”.  $\square$

## E Privacy Spheres

The aim of this appendix is to characterize privacy and inference in terms of spheres. Spheres exhibit homogeneity, which is good for privacy, while still admitting a coordinate system for identifiability.

We first prove a theorem characterizing individual attribute privacy, then a generalization that holds for arbitrary elements of a relation's poset, and finally a characterization of relations that preserve both attribute and association privacy.

### E.1 Individual Attribute Privacy

We first need a lemma as a tool. Recall also Definitions 6 and 65 (see pages 93 and 98).

**Lemma 69.** *Let  $R$  be a relation on  $X \times Y$ . Let  $x \in X$  be uniquely identifiable via  $R$ . Then:*

$$\left( \bigcap_{y \in Y_x} X_y \right) \setminus \{x\} = \emptyset.$$

*Moreover,  $R$  preserves attribute privacy for  $x$  if and only if*

$$\left( \bigcap_{y \in \gamma} X_y \right) \setminus \{x\} \neq \emptyset, \quad \text{for all } \gamma \subsetneq Y_x.$$

*Proof.* The first statement follows from the definition of unique identifiability:  $\bigcap_{y \in Y_x} X_y = \psi_R(Y_x) = \{x\}$ .

For the second statement:

I. Assume that  $R$  preserves attribute privacy for  $x$ . Let  $\gamma \subsetneq Y_x$ . If  $(\bigcap_{y \in \gamma} X_y) \setminus \{x\} = \emptyset$ , then  $\psi_R(\gamma) = \bigcap_{y \in \gamma} X_y = \{x\}$ , since  $x \in X_y$  whenever  $y \in \gamma \subseteq Y_x$  (when  $\gamma = \emptyset$ , the vacuous intersection is all of  $X$ , containing  $x$ ). That says a proper subset of  $Y_x$  identifies  $x$ , leading to a contradiction, as in the proof of Lemma 60 on page 93.

II. Assume  $(\bigcap_{y \in \gamma} X_y) \setminus \{x\} \neq \emptyset$  for all proper subsets  $\gamma$  of  $Y_x$ . If  $R$  fails to preserve attribute privacy for  $x$ , then by Lemma 66 there is some  $\gamma$  of the form  $Y_x \setminus \{y\}$ , with  $y \in Y_x$ , such that  $\gamma \subsetneq (\phi_R \circ \psi_R)(\gamma) = Y_x$ . Applying  $\psi_R$  to both sides of that last equality gives  $\psi_R(\gamma) = \psi_R(Y_x) = \{x\}$ , by unique identifiability. That is a contradiction, since  $\psi_R(\gamma) = \bigcap_{y \in \gamma} X_y$ .  $\square$

We now address our characterization of individual privacy, proving a theorem stated previously:

**Theorem 9** (Individual Attribute Privacy). *Let  $R$  be a relation on  $X \times Y$ , with  $|X| > 1$ . Suppose  $x \in X$  is uniquely identifiable via  $R$ . Let  $Q$  be the relation modeling  $\text{Lk}(\Psi_R, x)$ . Then the following three conditions are equivalent:*

- (a)  $R$  preserves attribute privacy for  $x$ ,
- (b)  $\text{Lk}(\Psi_R, x) \simeq \mathbb{S}^{k-2}$ , with  $k = |Y_x|$ ,
- (c)  $\Phi_Q = \partial(Y_x)$ .



*Proof.* The hypotheses ensure that  $Y_x \neq \emptyset$  (and so also  $Y \neq \emptyset$ ). They also ensure that  $x$  is a vertex of  $\Psi_R$ , so the link is not void. It could be the empty complex  $\{\emptyset\}$ , of course.

Observe that  $Q$  is the restriction of  $R$  to  $\overline{X} \times Y_x$ , with  $\overline{X} = \bigcup_{y \in Y_x} X_y \setminus \{x\}$ .

If  $\overline{X} = \emptyset$ , then, reasoning as in the proof of Lemma 54 on page 90, we see that  $\text{Lk}(\Psi_R, x) = \{\emptyset\} = \mathbb{S}^{-1}$ . Furthermore,  $x$  does not share any of its attributes with any other individuals in  $X$ . By convention,  $\Phi_Q = \{\emptyset\}$  as well. If  $k = |Y_x| = 1$ , meaning  $x$  has a single attribute, then  $R$  preserves attribute privacy for  $x$  since  $|X| > 1$ . Also,  $\mathbb{S}^{k-2} = \mathbb{S}^{-1} = \{\emptyset\} = \partial(Y_x)$ . So conditions (a), (b), (c) all hold. If  $k = |Y_x| \geq 2$ , then any one attribute of  $Y_x$  implies all the others, so condition (a) does not hold. Moreover, conditions (b) and (c) also do not hold. In short, the theorem holds when  $\overline{X} = \emptyset$ .

We now assume that  $\overline{X} \neq \emptyset$ . We then know that  $\text{Lk}(\Psi_R, x) = \Psi_Q \simeq \Phi_Q$  by a dual version of Lemma 54 and by Dowker duality. Definitionally,  $\partial(Y_x) \simeq \mathbb{S}^{k-2}$ , with  $k = |Y_x| > 0$ . We therefore see that (c) implies (b). To see that (b) implies (c), observe that the underlying vertex set of  $\Phi_Q$  is  $Y_x$ , so  $\Phi_Q \simeq \mathbb{S}^{k-2}$  means  $\Phi_Q = \partial(Y_x)$ , since no proper subset of a sphere can be homotopic to that same sphere. To prove the theorem we therefore only need to establish that conditions (a) and (c) are equivalent.

Recall the formulas relating  $\phi_Q$  and  $\phi_R$  from page 91 and dualize them here. We see that:

$$\psi_Q(\chi) = \psi_R(\chi) \setminus \{x\} = \left( \bigcap_{y \in \chi} X_y \right) \setminus \{x\}, \quad \text{for all } \emptyset \neq \chi \subseteq Y_x.$$

I. Assume that  $R$  preserves attribute privacy for  $x$ . By Lemma 69 and the formula above we see that  $\psi_Q(\chi) \neq \emptyset$  for all nonempty proper subsets  $\chi$  of  $Y_x$  and that  $\psi_Q(Y_x) = \emptyset$ , since  $x$  is uniquely identifiable. Consequently,  $\Phi_Q$  contains every nonempty proper subset of  $Y_x$  as a simplex, but does not contain  $Y_x$ . (Also,  $\Phi_Q$  contains the empty simplex since the complex is not void.) Thus  $\Phi_Q = \partial(Y_x)$ .

II. Assume that  $\Phi_Q = \partial(Y_x)$ . Then  $\psi_Q(\chi) \neq \emptyset$  for every nonempty proper subset  $\chi$  of  $Y_x$ . By the formula above,  $(\bigcap_{y \in \chi} X_y) \setminus \{x\} \neq \emptyset$ , for each such  $\chi$ . Now suppose  $\chi = \emptyset \subsetneq Y_x$ . Then:

$$\emptyset \neq \overline{X} = \psi_Q(\emptyset) \subseteq X \setminus \{x\} = \left( \bigcap_{y \in \emptyset} X_y \right) \setminus \{x\}.$$

So we see that  $(\bigcap_{y \in \gamma} X_y) \setminus \{x\} \neq \emptyset$  for every proper subset of  $Y_x$ , implying that  $R$  preserves attribute privacy for  $x$ , by Lemma 69.  $\square$

Comment: It is impossible to satisfy the following three conditions simultaneously:

(1)  $x$  is uniquely identifiable, (2)  $|Y_x| = 1$ , (3)  $\overline{X} \neq \emptyset$ .

## E.2 Group Attribute Privacy

We now generalize the previous theorem to arbitrary elements  $(\sigma, \gamma)$  of the doubly-labeled poset  $P_R$  associated with a relation  $R$ . We stated the generalized theorem previously in the report, as Theorem 10, and replicate that below. One may view this generalized theorem as a characterization of the conditions under which a *group*  $\sigma$  of individuals has its attribute privacy preserved, as a whole not necessarily individually. Theorem 9 is a special case of Theorem 10, with the “group” a single individual  $x$ , since  $(\{x\}, Y_x) \in P_R$  whenever  $x$  is uniquely identifiable via  $R$ .

**Theorem 10** (Group Attribute Privacy). *Let  $R$  be a relation on  $X \times Y$ . Suppose  $(\sigma, \gamma) \in P_R$ , with  $\sigma \neq X$ . Let  $Q$  be the relation modeling  $\text{Lk}(\Psi_R, \sigma)$ . Then the following three conditions are equivalent:*

- (a)  $(\phi_R \circ \psi_R)(\gamma') = \gamma'$ , for every subset  $\gamma'$  of  $\gamma$ ,
- (b)  $\text{Lk}(\Psi_R, \sigma) \simeq \mathbb{S}^{k-2}$ , with  $k = |\gamma|$ ,
- (c)  $\Phi_Q = \partial(\gamma)$ .

*Proof.* Reminder: Since  $(\sigma, \gamma) \in P_R$ ,  $\emptyset \neq \sigma \in \Psi_R$ ,  $\emptyset \neq \gamma \in \Phi_R$ ,  $\phi_R(\sigma) = \gamma$ , and  $\psi_R(\gamma) = \sigma$ .

Thus also  $(\phi_R \circ \psi_R)(\gamma) = \gamma$ , meaning we can focus on proper subsets of  $\gamma$  for part (a).

Recall also that  $Q$  is the restriction of  $R$  to  $\overline{X} \times \gamma$ , with  $\overline{X} = \bigcup_{y \in \gamma} X_y \setminus \sigma$ .

If  $\overline{X} = \emptyset$ , then  $\text{Lk}(\Psi_R, \sigma) = \{\emptyset\} = \mathbb{S}^{-1}$ . By convention,  $\Phi_Q = \{\emptyset\}$  as well. If  $k = |\gamma| = 1$ , then  $\mathbb{S}^{k-2} = \mathbb{S}^{-1} = \{\emptyset\} = \partial(\gamma)$ . The only proper subset of  $\gamma$  in this case is  $\gamma' = \emptyset$ , and  $(\phi_R \circ \psi_R)(\emptyset) = \phi_R(X) = \emptyset$ . (Reason: If  $y \in \phi_R(X)$ , then  $y \in \gamma$ , so  $\gamma = \{y\}$ , implying  $\sigma = X$ , which is disallowed.) Thus conditions (a), (b), (c) all hold. If  $k = |\gamma| \geq 2$ , then conditions (b) and (c) cannot hold. Also, condition (a) does not hold since  $(\phi_R \circ \psi_R)(\{y\}) = \gamma$  for each  $y \in \gamma$ , bearing in mind that  $\overline{X} = \emptyset$  means  $X_y = \sigma$  for each  $y \in \gamma$ . In short, the theorem holds when  $\overline{X} = \emptyset$ .

We now assume that  $\overline{X} \neq \emptyset$ . As in the proof of Theorem 9, we see readily that conditions (b) and (c) are equivalent, so we will prove that conditions (a) and (c) are equivalent. And, as in the previous proof, dualizing a formula from page 91 gives this formula:

$$\psi_Q(\chi) = \psi_R(\chi) \setminus \sigma, \quad \text{for all } \emptyset \neq \chi \subseteq \gamma.$$

I. Assume that  $(\phi_R \circ \psi_R)(\gamma') = \gamma'$ , for every subset  $\gamma'$  of  $\gamma$ .

We will establish that  $\Phi_Q$  contains all proper subsets of  $\gamma$  but not  $\gamma$ , telling us  $\Phi_Q = \partial(\gamma)$ . Since  $\Phi_Q$  is not void, it contains the empty simplex.

Pick some  $\emptyset \neq \gamma' \subsetneq \gamma$ . Since  $(\phi_R \circ \psi_R)(\gamma') = \gamma'$ ,  $\psi_R(\gamma') \supsetneq \sigma$ .

The formula above therefore says  $\psi_Q(\gamma') \neq \emptyset$ , telling us  $\gamma' \in \Phi_Q$ .

Similarly,  $\psi_Q(\gamma) = \psi_R(\gamma) \setminus \sigma = \sigma \setminus \sigma = \emptyset$ , so  $\gamma \notin \Phi_Q$ .

II. Assume that  $\Phi_Q = \partial(\gamma)$ .

Recall that  $k = |\gamma| > 0$ . We look at two cases based on the value of  $k$ :

$k = 1$ : In this case,  $\gamma = \{y\}$ , for some  $y \in Y$ , so  $\sigma = X_y$  and  $\overline{X} = \emptyset$ , which we discussed above.

$k > 1$ : Suppose, for the sake of contradiction, that  $\gamma' \subsetneq (\phi_R \circ \psi_R)(\gamma')$ , for some  $\gamma' \subsetneq \gamma$ . By Lemma 47 on page 88, we can assume  $\gamma' = \gamma \setminus \{y\}$ , for some  $y \in \gamma$ . Consequently,  $(\phi_R \circ \psi_R)(\gamma') = \gamma$ , which implies  $\psi_R(\gamma') = \sigma$ . The formula above then says  $\psi_Q(\gamma') = \emptyset$ , whereas the fact that  $\gamma' \in \Phi_Q$  means  $\psi_Q(\gamma') \neq \emptyset$ , a contradiction.  $\square$

The following lemma, previously stated on page 29, relates privacy preservation in a link to privacy preservation in the encompassing relation.

**Lemma 11** (Interpreting Local Operators). *Let  $R$  be a relation on  $X \times Y$ . Suppose  $(\sigma, \gamma) \in P_R$ , with  $\sigma \neq X$ . Let  $Q$  be the relation on  $\overline{X} \times \gamma$  that models  $\text{Lk}(\Psi_R, \sigma)$  and suppose  $\overline{X} \neq \emptyset$ .*

*Then, for every  $\gamma' \subseteq \gamma$ :*

*(i) If  $\gamma' \notin \Phi_Q$ , then  $\psi_R(\gamma') = \sigma$ ,*

*(ii) If  $\gamma' \in \Phi_Q$ , then  $\psi_R(\gamma') \supsetneq \sigma$ .*

*Moreover, in this case:*

*If  $(\phi_Q \circ \psi_Q)(\emptyset) = \emptyset$ , then  $(\phi_R \circ \psi_R)(\emptyset) = \emptyset$ .*

*If  $\gamma' \neq \emptyset$ , then  $(\phi_Q \circ \psi_Q)(\gamma') = (\phi_R \circ \psi_R)(\gamma')$ .*

*Proof.* Observe that for every  $\gamma' \subseteq \gamma$ , one has  $\gamma' \in \Phi_R$  and  $\psi_R(\gamma') \supseteq \psi_R(\gamma) = \sigma$ .

By the formula on page 91 dualized, if  $\emptyset \neq \gamma' \subseteq \gamma$ , then  $\psi_Q(\gamma') = \psi_R(\gamma') \setminus \sigma$ .

(i) Suppose  $\gamma' \notin \Phi_Q$ . Then  $\gamma' \neq \emptyset$  since  $\emptyset \in \Phi_Q$ . Also,  $\psi_Q(\gamma') = \emptyset$ , so by the formula above,  $\psi_R(\gamma') = \sigma$ .

(ii) Suppose  $\gamma' \in \Phi_Q$ . If  $\gamma' = \emptyset$ , then  $\psi_R(\emptyset) = X \supsetneq \sigma$ , by hypothesis. If  $\gamma' \neq \emptyset$ , then  $\psi_Q(\gamma') \neq \emptyset$ , so again by the formula above,  $\psi_R(\gamma') \supsetneq \sigma$ .

Turning to the “Moreover”:

If  $y \in (\phi_R \circ \psi_R)(\emptyset)$ , then  $y$  is an attribute for all individuals in  $X$ , so  $y \in \gamma$  and  $y \in \phi_Q(\overline{X}) = (\phi_Q \circ \psi_Q)(\emptyset)$ .

Let  $\emptyset \neq \gamma' \in \Phi_Q$ . By the formula on page 90 dualized, if  $\kappa \subseteq \overline{X}$ , then  $\phi_Q(\kappa) = \phi_R(\kappa \cup \sigma)$ .

Therefore:  $(\phi_Q \circ \psi_Q)(\gamma') = \phi_Q(\psi_R(\gamma') \setminus \sigma) = (\phi_R \circ \psi_R)(\gamma')$ .  $\square$

**Comment:** Also,  $(\phi_Q \circ \psi_Q)(\emptyset) = \phi_Q(\overline{X}) = \phi_R(\bigcup_{y \in \gamma} X_y) = \bigcap_{y \in \gamma} (\phi_R \circ \psi_R)(\{y\})$ .

### E.3 Preserving Attribute and Association Privacy

In this subsection, we are interested in understanding relations that preserve *both* attribute and association privacy. We will discover that this requirement is severely limiting. As one can already see from Theorem 64 on page 97, if  $R$  is a nonvoid tight relation on  $X \times Y$  that preserves both attribute and association privacy, then  $|X| = |Y| = n$ . What are the possibilities?

$n = 0$ : Not possible; this is the void relation.

$n = 1$ : Not possible; such a relation does not preserve privacy; one can infer the single individual or single attribute from nothing.

$n = 2$ : As we have seen before, such a relation must be isomorphic to the following relation:

$R$	$y_1$	$y_2$
$x_1$	•	
$x_2$		•

Then both  $\Psi_R$  and  $\Phi_R$  are instances of the 0-sphere  $\mathbb{S}^0$ .

$n \geq 3$ : Now there are several possibilities:

- The relation could be isomorphic to a *cyclic staircase relation*:

$R$	$y_1$	$y_2$	$\cdots$	$\cdots$	$y_{n-1}$	$y_n$
$x_1$	•	•				
$x_2$		•	•			
$\vdots$			$\ddots$	$\ddots$		
$\vdots$				$\ddots$	•	
$x_{n-1}$					•	•
$x_n$	•					•

Then both  $\Psi_R$  and  $\Phi_R$  are homotopic to the 1-sphere  $\mathbb{S}^1$ . Each is simply a linear cycle of edges, with vertices in one complex dualizing to edges in the other.

- The relation could be isomorphic to a *spherical boundary relation* in which every entry is present except that a diagonal is blank. For example, in the following relation all entries are present except those for  $(x_i, y_{n-i+1})$ ,  $i = 1, \dots, n$ :

$R$	$y_1$	$y_2$	$\cdots$	$\cdots$	$y_{n-1}$	$y_n$
$x_1$	•	•	•	$\cdots$	•	
$x_2$	•	•	$\cdots$	•		•
$\vdots$	•	$\vdots$	•		•	•
$\vdots$	•	•		•	$\vdots$	•
$x_{n-1}$	•		•	$\cdots$	•	•
$x_n$		•	$\cdots$	•	•	•

Then  $\Psi_R$  and  $\Phi_R$  are each boundary complexes, namely  $\Psi_R = \partial(X)$  and  $\Phi_R = \partial(Y)$ . Thus both are homotopic to the  $(n - 2)$ -sphere  $\mathbb{S}^{n-2}$ .

- Finally,  $R$  could have multiple components, each of which is one of the following: A singleton, a cyclic staircase relation, or a spherical boundary relation, all as above. (Observe that even though a  $1 \times 1$  relation in and of itself preserves no privacy, a relation can preserve privacy over a  $1 \times 1$  subrelation when that subrelation is one of several components.)

(Comment: the staircase and spherical relations are isomorphic when  $n = 3$ .)

The aim of this subsection is to prove that these are the only possibilities.

**Lemma 70.** *Let  $R$  be a connected tight relation on  $X \times Y$ , with  $|X| = |Y| \geq 3$ , that preserves both attribute and association privacy.*

*Let  $x \in X$  and define  $Q$  to be the relation on  $\overline{X} \times Y_x$  that models  $\text{Lk}(\Psi_R, x)$ .*

*Then  $\Psi_Q = \partial(\overline{X})$  and  $\Phi_Q = \partial(Y_x)$ , with  $|\overline{X}| = |Y_x|$ .*

*Proof.* Observe that  $Y_x \neq \emptyset$  since  $R$  is tight. Recall that  $\overline{X} = \bigcup_{y \in Y_x} X_y \setminus \{x\}$ , which is nonempty since  $R$  is connected and  $X$  contains not just  $x$ .

By Lemma 63 on page 95,  $x$  is uniquely identifiable via  $R$ , so Theorem 9 on page 100 says that  $\Psi_Q \simeq \mathbb{S}^{k-2}$  and  $\Phi_Q = \partial(Y_x)$ , with  $k = |Y_x|$ . If we can show that  $|\overline{X}| = k$ , then we can conclude that  $\Psi_Q = \partial(\overline{X})$ .

The vertices of  $\Psi_Q$  generate the maximal simplices of  $\Phi_Q$ . In particular, there exist  $x_1, \dots, x_k \in \overline{X}$  such that  $\overline{Y}_1, \dots, \overline{Y}_k$  are the maximal simplices of  $\Phi_Q$ , with  $\overline{Y}_i = Y_{x_i} \cap Y_x$ , and  $|\overline{Y}_i| = k - 1$ , for  $i = 1, \dots, k$ .

Let  $\tilde{x} \in \overline{X}$ . Then  $Y_{\tilde{x}} \cap Y_x \subseteq \overline{Y}_i \subseteq Y_{x_i}$ , for some  $i \in \{1, \dots, k\}$ .

That says  $\emptyset \neq \psi_R(\{\tilde{x}, x\}) \subseteq \psi_R(\{x_i\})$ .

Since  $R$  preserves association privacy, the dualization of Lemma 59 on page 93 implies  $\tilde{x} = x_i$ . Thus  $|\overline{X}| = k$ .  $\square$

Comment: Where did we use the assumption that each of  $X$  and  $Y$  has at least three elements? In fact, for the proof it is enough to assume that  $|X| = |Y| \geq 2$ . However, there is no connected tight relation that preserves privacy when  $|X| = |Y| = 2$ .

**Corollary 71.** *Let  $R$  be a connected tight relation on  $X \times Y$ , with  $|X| = |Y|$ , that preserves both attribute and association privacy.*

*Let  $y \in Y$  and suppose  $|X_y| \geq 4$ .*

*Then  $\text{Lk}(\Phi_R, y)$  is not a linear cycle. (In other words, the relation  $Q$  that models  $\text{Lk}(\Phi_R, y)$  is not isomorphic to a staircase relation.)*

*Proof.* Arguing as in the proof of Lemma 70, now in dual form, we see that  $\text{Lk}(\Phi_R, y) \simeq \mathbb{S}^{k-2}$ , with  $k = |X_y|$ . Since  $k - 2 \geq 2$ ,  $\text{Lk}(\Phi_R, y)$  is not a linear cycle.  $\square$

**Corollary 72.** *Let  $R$  be a connected tight relation on  $X \times Y$ , with  $|X| = |Y| \geq 3$ , that preserves both attribute and association privacy.*

*Suppose  $\{x, x'\}$ , with  $x \neq x'$ , is an edge (1-simplex) in  $\Psi_R$ .*

*Then  $|Y_x| = |Y_{x'}|$ .*

*Proof.* Let  $k = |Y_x|$  and  $k' = |Y_{x'}|$ .

Observe that  $x'$  is a vertex of  $\text{Lk}(\Psi_R, x)$  and  $x$  is a vertex of  $\text{Lk}(\Psi_R, x')$ .

By the proof of Lemma 70, each of  $x'$  and  $x$  generates a maximal simplex in the attribute complex associated with the other's link. That simplex is  $Y_x \cap Y_{x'}$  in both complexes.

So  $k - 1 = |Y_x \cap Y_{x'}| = k' - 1$ , hence  $k = k'$ .  $\square$

**Corollary 73.** *Let  $R$  be a connected tight relation on  $X \times Y$ , with  $|X| = |Y| \geq 3$ , that preserves both attribute and association privacy.*

*Then all rows and columns have the same number of nonblank entries.*

*Proof.* By Lemma 50 on page 88 and Corollary 72 above, all rows have the same number  $k_r$  of nonblank entries. Dualizing, one sees that all columns have the same number  $k_c$  of nonblank entries. We claim that  $k_c = k_r$ . This assertion follows from Lemma 70 and its proof as follows:

Pick some  $x \in X$  and let  $Q$  be the relation modeling  $\text{Lk}(\Psi_R, x)$ . By Lemma 70,  $\Psi_Q$  and  $\Phi_Q$  are each boundary complexes, with  $k_r = |Y_x|$  vertices. Moreover, each element  $y \in Y_x$  generates a maximal simplex  $X_y \cap \overline{X}$  in  $\Psi_Q$ , which must have size  $k_r - 1$ . The column  $X_y$  contains one additional element, namely  $x$ . So  $k_c = |X_y| = (k_r - 1) + 1 = k_r$ .  $\square$

**Theorem 74** (Privacy as Sphere). *Let  $R$  be a nonvoid connected tight relation on  $X \times Y$  that preserves both attribute and association privacy.*

*Then  $|X| = |Y| \geq 3$  and  $R$  is isomorphic to either a cyclic staircase relation or a spherical boundary relation (each described on page 104).*

*Proof.* As we commented previously, Theorem 64 implies that  $|X| = |Y| = n$ . Connectedness further means that  $n \geq 3$ .

By Corollary 73, all rows and columns in  $R$  have the same number of nonblank entries. In other words,  $|X_y| = |Y_x| = k$ , for all  $x \in X$  and all  $y \in Y$ , for some fixed  $k$ . By connectedness,  $k \geq 2$ .

By Lemma 63 on page 95, each  $x \in X$  is uniquely identifiable via  $R$ . Dualized, each  $y \in Y$  is uniquely identifiable via  $R$  as well.

If  $k = 2$ , then  $\Psi_R$  and  $\Phi_R$  contain vertices and edges but no higher-dimensional simplices. By duality, each vertex therefore has at most two incident edges. By unique identifiability, each vertex has exactly two incident edges. Thus, by connectedness, each complex is a linear cycle. So  $R$  is isomorphic to a staircase relation.

Now assume that  $k \geq 3$ .

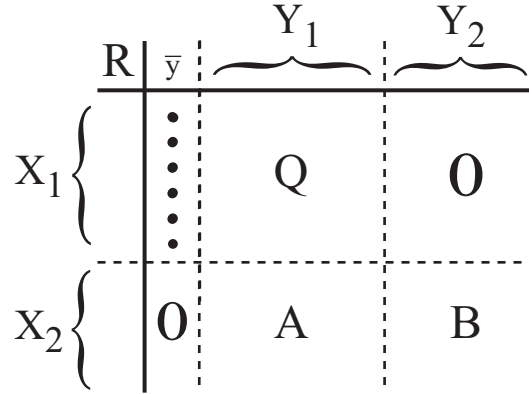
Pick a  $\overline{y} \in Y$  and consider the decomposition of Figure 52, similar to the one we saw in the proof of Lemma 63. (We indicate blank entries either by blanks or by explicit “0”s.)

Let  $X_1 = X_{\overline{y}}$  and write  $X = X_1 \cup X_2$  with  $X_2 = X \setminus X_1$ .  $X_1 \neq \emptyset$  since every column of  $R$  has  $k$  nonblank entries and  $X_2 \neq \emptyset$  since  $R$  preserves attribute privacy.

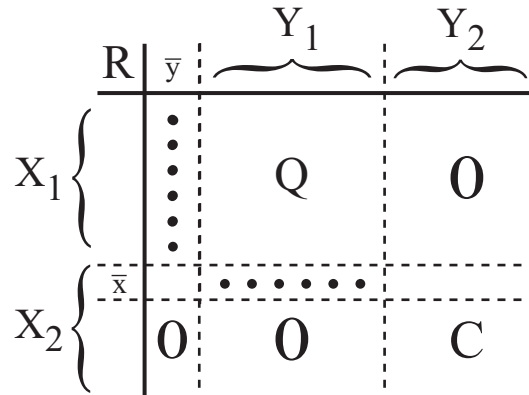
Let  $Q$  model  $\text{Lk}(\Phi_R, \overline{y})$ . So  $Q$  is  $R$  restricted to  $X_1 \times Y_1$ , with  $Y_1 = \bigcup_{x \in X_1} Y_x \setminus \{\overline{y}\}$ .  $Y_1 \neq \emptyset$  because every row of  $R$  has  $k$  nonblank entries. In particular, there are exactly  $k - 1$  entries in each row of  $Q$ , so at least two entries in each row.

Now write  $Y$  as the disjoint union  $Y = \{\overline{y}\} \cup Y_1 \cup Y_2$ , with  $Y_2 = Y \setminus (Y_1 \cup \{\overline{y}\})$ . Observe that every element in  $X_1$  has attribute  $\overline{y}$  but has no attributes in  $Y_2$ , by construction.

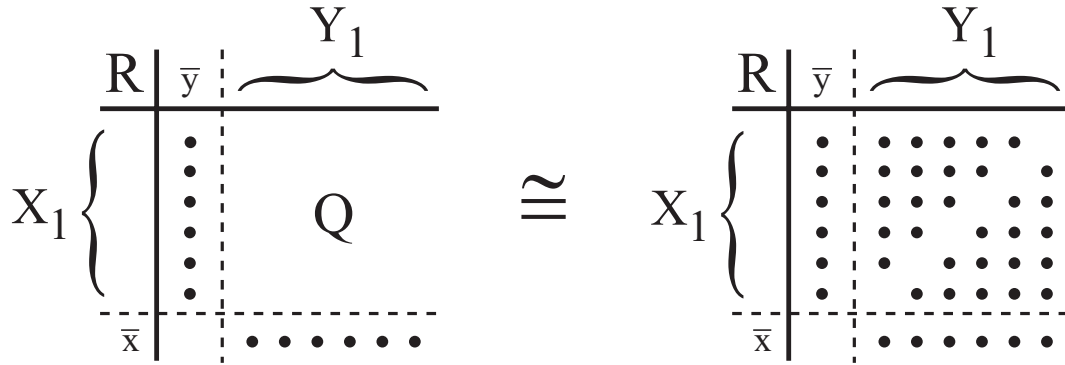
By the dual to Lemma 70, we know that  $\Psi_Q = \partial(X_1)$  and  $\Phi_Q = \partial(Y_1)$ , with  $k = |X_1| = |Y_1|$ . Therefore, for each  $y \in Y_1$ , column  $X_y$  of  $R$  has  $k - 1$  entries that lie in  $X_1$  and one entry

Figure 52: Relation  $R$  decomposed into blocks for the proof of Theorem 74.

that lies in  $X_2$ . We claim that the  $X_2$  entry is the same across all columns  $X_y$  as  $y$  varies over  $Y_1$ . For otherwise, at least two such columns would have an intersection (nonempty, since  $k - 2 \geq 1$ ) contained wholly within  $X_{\bar{y}}$ , implying that  $R$  permits attribute inference after all, by the proof of Lemma 59 on page 93. Call that common element  $\bar{x}$ . Observe that  $Y_{\bar{x}} = Y_1$  since every row of  $R$  has exactly  $k$  elements. Consequently, the block diagram for  $R$  becomes as in Figure 53.

Figure 53: Relation  $R$  decomposed further.

Observe that no element of  $X_1 \cup \{\bar{x}\}$  has any attributes in  $Y_2$  and that no element of  $X_2 \setminus \{\bar{x}\}$  has any attributes in  $Y_1 \cup \{\bar{y}\}$ , by the row and column cardinality constraints. That means relation  $C$ , which is the restriction of  $R$  to  $(X_2 \setminus \{\bar{x}\}) \times Y_2$  is disconnected from the rest of  $R$ , if  $C$  were to exist. We conclude that  $Y_2 = \emptyset$  and that  $X_2 = \{\bar{x}\}$ . Thus, finally,  $R$  must decompose as in Figure 54. As we have seen,  $Q$  is nearly a full relation, missing a diagonal. We now see that  $R$  is also nearly a full relation, missing a diagonal. Thus  $\Psi_R = \partial(X)$  and  $\Phi_R = \partial(Y)$ , meaning  $R$  is a spherical boundary relation, as claimed.  $\square$

Figure 54: Relation  $R$  decomposes diagonally.

**Corollary 75.** *Let  $R$  be a nonvoid tight relation that preserves both attribute and association privacy. Decompose  $R$  into its connected components as  $R = R_1 \cup \dots \cup R_\ell$ , with  $R_i$  a relation on  $X_i \times Y_i$ , as per the proof of Lemma 51 on page 89. Then  $R_i$  is either a singleton or a staircase relation or a spherical boundary relation and  $|X_i| = |Y_i|$ ,  $i = 1, \dots, \ell$ .*

Comment: When  $\ell = 2$  and each of  $R_1$  and  $R_2$  is a singleton, then the Dowker complexes of  $R$  itself,  $\Psi_R$  and  $\Phi_R$ , are each an instance of  $\mathbb{S}^0$ .

*Proof.* Consider some  $R_i$ .

Suppose that  $X_i \in \Psi_{R_i}$ . Then some attribute  $y \in Y_i$  is shared by all individuals in  $X_i$ . If there were any other attributes in  $Y_i$ , then each of those would individually imply  $y$  in  $R$ . Since  $R$  preserves attribute privacy,  $|Y_i| = 1$ . Consequently, since  $R$  also preserves association privacy,  $|X_i| = 1$ , so  $R_i$  is a singleton.

If  $R_i$  is not a singleton, then  $X_i \notin \Psi_{R_i}$  and similarly  $Y_i \notin \Phi_{R_i}$ .

Consequently, Lemma 51 and Corollary 53 on page 89 tell us that  $R_i$  is a nonvoid connected tight relation that preserves both attribute and association privacy. Theorem 74 completes the proof.  $\square$



## F Poset Chains

Recall Definition 12, on page 38, of the Galois lattice  $P_R^+$  associated with a relation  $R$ , and Definition 13, on page 41, defining informative attribute release sequences. In this appendix we will explore connections between these two concepts.

### F.1 Maximal Chains and Informative Attribute Release Sequences

Let  $R$  be a nonvoid relation on  $X \times Y$ . Suppose  $\{(\sigma_k, \gamma_k) < \cdots < (\sigma_1, \gamma_1) < (\sigma_0, \gamma_0)\}$  is a maximal chain in  $P_R^+$ . Then, for  $1 \leq i \leq k$ ,  $\sigma_i \subsetneq \sigma_{i-1}$  and  $\gamma_i \supsetneq \gamma_{i-1}$ .

Also,  $\sigma_0 = X$  and  $\gamma_k = Y$ . Note that  $\gamma_0 = \phi_R(X)$  and  $\sigma_k = \psi_R(Y)$ . Consequently,  $\gamma_0 \neq \emptyset$  if and only if  $X \in \Psi_R$ , and  $\sigma_k \neq \emptyset$  if and only if  $Y \in \Phi_R$ .

We sometimes speak of a *maximal chain at and above*  $(\sigma, \gamma)$ , by which we mean a chain  $\{(\sigma, \gamma) < \cdots < (\sigma_1, \gamma_1) < (\sigma_0, \gamma_0)\}$  in  $P_R^+$  that is maximal among all such chains. Such a chain is a prefix of a full maximal chain in  $P_R^+$  (“prefix” with respect to our subscript ordering, which starts at the top of a poset and moves downward).

**Lemma 20** (Informative Attributes from Maximal Chains). *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. Suppose  $\{(\sigma_k, \gamma_k) < \cdots < (\sigma_1, \gamma_1) < (\sigma_0, \gamma_0)\}$ , with  $k \geq 1$ , is a maximal chain in  $P_R^+$ .*

*Define  $y_1, \dots, y_k$  by selecting some  $y_i \in \gamma_i \setminus \gamma_{i-1}$ , for each  $i = 1, \dots, k$ .*

*Then  $y_1, \dots, y_k$  is an informative attribute release sequence for  $R$ .*

*Moreover,  $(\phi_R \circ \psi_R)(\{y_1, \dots, y_i\}) = \gamma_i$  for each  $i = 0, 1, \dots, k$ .*

*Proof.* Establishing the “Moreover” also establishes the “iars” assertion.

The proof is by induction on  $i$ .

For the base case,  $i = 0$  and we need to show that  $(\phi_R \circ \psi_R)(\emptyset) = \gamma_0$ .

Calculating,  $(\phi_R \circ \psi_R)(\emptyset) = \phi_R(X) = \gamma_0$ , by our earlier comments about maximal chains.

For the induction step, we assume that, for some  $1 \leq i \leq k$ , the assertion holds for indices smaller than  $i$  and we need to show the assertion holds for  $i$ . First, observe:

$$\psi_R(\{y_1, \dots, y_i\}) = \psi_R(\{y_1, \dots, y_{i-1}\}) \cap X_{y_i} = \psi_R(\gamma_{i-1}) \cap X_{y_i} = \psi_R(\gamma_{i-1} \cup \{y_i\}).$$

(The middle equality follows from the induction hypothesis and a dual version of Corollary 45 from page 87, specifically because  $(\phi_R \circ \psi_R)(\{y_1, \dots, y_{i-1}\}) = \gamma_{i-1}$  and  $\psi_R \circ \phi_R \circ \psi_R = \psi_R$ .)

Since  $\gamma_{i-1} \subsetneq \gamma_{i-1} \cup \{y_i\} \subseteq \gamma_i$ ,

$$\gamma_{i-1} = (\phi_R \circ \psi_R)(\gamma_{i-1}) \subsetneq (\phi_R \circ \psi_R)(\gamma_{i-1} \cup \{y_i\}) \subseteq (\phi_R \circ \psi_R)(\gamma_i) = \gamma_i,$$

By maximality of the original chain and the nature of elements in  $P_R^+$ , we see that  $(\phi_R \circ \psi_R)(\gamma_{i-1} \cup \{y_i\}) = \gamma_i$ , so  $(\phi_R \circ \psi_R)(\{y_1, \dots, y_i\}) = (\phi_R \circ \psi_R)(\gamma_{i-1} \cup \{y_i\}) = \gamma_i$ .  $\square$

Here is a partial converse:

**Lemma 21** (Chains from Informative Attributes). *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty. Suppose  $y_1, \dots, y_k$  is an informative attribute release sequence for  $R$ , with  $k \geq 1$ .*

*Let  $\gamma_i = (\phi_R \circ \psi_R)(\{y_1, \dots, y_i\})$  and  $\sigma_i = \psi_R(\gamma_i)$ , for  $i = 1, \dots, k$ .*

*Let  $\gamma_0 = \phi_R(X)$ . Then  $\{(\sigma_k, \gamma_k) < \dots < (\sigma_1, \gamma_1) < (X, \gamma_0)\}$  is a chain in  $P_R^+$ .*

Comment: The resulting chain need not be maximal.

*Proof.* Observe that each  $(\sigma_i, \gamma_i) \in P_R^+$  by construction, so we need to establish the total ordering. Let's define  $\sigma_0 = X$ . We need to show that  $\sigma_i \subsetneq \sigma_{i-1}$ , for each  $i = 1, \dots, k$ .

Since  $\{y_1, \dots, y_i\} \supseteq \{y_1, \dots, y_{i-1}\}$ , we see that  $\sigma_i \subseteq \sigma_{i-1}$ . If  $\sigma_i = \sigma_{i-1}$ , then also  $\gamma_i = \gamma_{i-1}$ , contradicting the fact that  $y_i \in \gamma_i \setminus \gamma_{i-1}$  (which is true by the nature of informative attribute release sequences).  $\square$

As a corollary to Lemmas 20 and 21, one sees that every informative attribute release sequence (iars) for  $R$  is a subsequence of an iars derived from a maximal chain in  $P_R^+$ . (Technically, one needs to show that any nonempty subsequence of an iars is itself an iars. And one needs to show that extending any chain obtained via Lemma 21 to a maximal chain retains the original iars as a subsequence of one subsequently obtainable via Lemma 20. All that is straightforward.)

## F.2 Chains and Links

We are interested in understanding how chains and informative attribute release sequences behave as one passes to links. (Small caution: whereas we were looking at chains in  $P_R^+$  before, we focus here on  $P_R$  (and  $P_Q$ ).)

**Lemma 76** (Chains in Links). *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty and suppose  $(\sigma, \gamma) \in P_R$ . Let  $Q$  be the relation modeling  $\text{Lk}(\Psi_R, \sigma)$ . Then*

$$P_Q = \{(\sigma' \setminus \sigma, \gamma') \mid (\sigma, \gamma) < (\sigma', \gamma') \in P_R\}.$$

Comments:

- $Q$  is the restriction of  $R$  to  $\overline{X} \times \gamma$ , with  $\overline{X} = \bigcup_{y \in \gamma} X_y \setminus \sigma$ .
- $P_Q$  could be empty. This occurs precisely when  $(\sigma, \gamma)$  is a maximal element of  $P_R$ , which occurs precisely when  $\text{Lk}(\Psi_R, \sigma) = \{\emptyset\}$ .
- If  $\sigma = X$ , then  $\text{Lk}(\Psi_R, \sigma) = \{\emptyset\}$  and so  $P_Q = \emptyset$ , given Definition 8 on page 27. Note however that Definition 18 of  $Q(\sigma, \gamma)$  on page 43 would mean that  $P_{Q(\sigma, \gamma)}$  is degenerate.
- $P_Q$  never contains the element  $\hat{0}_Q = (\emptyset, \gamma)$  of  $P_Q^+$ . That element corresponds to  $(\sigma, \gamma)$  in  $P_R$ , consistent with the idea of Lemma 11 on page 29 that one has “localized (to)  $\sigma$  upon observing  $\gamma$ ”. (See also Definition 15 on page 42.)

- $P_Q$  could contain the element  $\hat{1}_Q = (\bar{X}, \chi)$  of  $P_Q^+$ , for some  $\chi \subsetneq \gamma$ . That happens precisely when  $\bar{X} \neq \emptyset$  and all individuals in  $\bar{X}$  share an attribute of  $\gamma$ , in which case  $\chi \neq \emptyset$ .

*Proof.* The proof relies on dual versions of the formulas on pages 90–91.

I. Suppose  $(\kappa, \eta) \in P_Q$ . So  $\kappa \neq \emptyset$  and  $\eta \neq \emptyset$ . Also,  $\Psi_Q = \text{Lk}(\Psi_R, \sigma)$ , so  $\kappa \cap \sigma = \emptyset$  and  $\kappa \cup \sigma \in \Psi_R$ . Let  $\sigma' = \kappa \cup \sigma$ . So  $\sigma \subsetneq \sigma'$ . We can take  $\gamma'$  to be  $\eta$  since  $\eta = \phi_Q(\kappa) = \phi_R(\sigma')$ . Note that  $\psi_R(\gamma') = \psi_Q(\eta) \cup \sigma = \kappa \cup \sigma = \sigma'$ . We have shown that  $(\sigma', \gamma') \in P_R$  and  $(\sigma, \gamma) < (\sigma', \gamma')$ .

II. Suppose  $(\sigma', \gamma') \in P_R$  and  $(\sigma, \gamma) < (\sigma', \gamma')$ . So  $\sigma \subsetneq \sigma'$  and  $\gamma \supsetneq \gamma'$ . Let  $\kappa = \sigma' \setminus \sigma$ . Note that  $\kappa \neq \emptyset$  and  $\gamma' \neq \emptyset$ . Moreover,  $\kappa \in \text{Lk}(\Psi_R, \sigma)$ , so  $\bar{X} \neq \emptyset$ .

Verifying correspondence:  $\phi_Q(\kappa) = \phi_R(\sigma') = \gamma'$  and  $\psi_Q(\gamma') = \psi_R(\gamma') \setminus \sigma = \sigma' \setminus \sigma = \kappa$ .

We have shown that  $(\sigma' \setminus \sigma, \gamma') \in P_Q$ .  $\square$

**Corollary 77** (Order Preservation). *Let  $R$  and  $Q$  be as in Lemma 76, with  $(\sigma, \gamma) \in P_R$ .*

*Then  $(\sigma, \gamma) < (\sigma_1, \gamma_1) < (\sigma_2, \gamma_2)$  in  $P_R$  if and only if  $(\sigma_1 \setminus \sigma, \gamma_1) < (\sigma_2 \setminus \sigma, \gamma_2)$  in  $P_Q$ .*

*Proof.* By Lemma 76 and because:

(a)  $\sigma \subsetneq \sigma_1 \subsetneq \sigma_2$  implies  $\emptyset \neq \sigma_1 \setminus \sigma \subsetneq \sigma_2 \setminus \sigma$ ;

(b)  $\emptyset \neq \kappa_1 \subsetneq \kappa_2$  implies  $\sigma \subsetneq (\kappa_1 \cup \sigma) \subsetneq (\kappa_2 \cup \sigma)$ .  $\square$

**Corollary 78** (Maximal Chain Preservation). *Let  $R$  and  $Q$  be as in Lemma 76, with  $(\sigma, \gamma) \in P_R$ . Then  $\{(\sigma, \gamma) < (\sigma_k, \gamma_k) < \dots < (\sigma_1, \gamma_1)\}$  is a maximal chain at and above  $(\sigma, \gamma)$  in  $P_R$  if and only if  $\{(\sigma_k \setminus \sigma, \gamma_k) < \dots < (\sigma_1 \setminus \sigma, \gamma_1)\}$  is a maximal chain in  $P_Q$ .*

*Proof.* By Lemma 76 and Corollary 77, we know that  $\{(\sigma, \gamma) < (\sigma_k, \gamma_k) < \dots < (\sigma_1, \gamma_1)\}$  is a chain extending upward from  $(\sigma, \gamma)$  in  $P_R$  if and only if  $\{(\sigma_k \setminus \sigma, \gamma_k) < \dots < (\sigma_1 \setminus \sigma, \gamma_1)\}$  is a chain in  $P_Q$ .

Maximality follows for the same reason: Refine or extend a chain in one poset and one can refine or extend the corresponding chain in the other poset as well.  $\square$

**Comment about “length”:** Recall that the length of a chain in a poset is one less than the number of elements in the chain. We also speak of the *length* of an informative attribute release sequence  $y_1, \dots, y_k$ , which is  $k$ , the actual number of elements in the sequence.

In the context of Lemmas 20 and 21, there is a happy alignment of definitions: The length  $k$  of a longest iars in  $R$  is the length  $\ell(P_R^+)$ .

In thinking about poset lengths, bear in mind that  $\ell(P_R)$  may be any of  $\ell(P_R^+)$ ,  $\ell(P_R^+) + 1$ , or  $\ell(P_R^+) + 2$ , depending on whether the top and/or bottom elements of  $P_R^+$  already lie in  $P_R$ .

**Corollary 79** (Longest Localization Sequences). *Let  $R$  be a relation on  $X \times Y$ , with both  $X$  and  $Y$  nonempty and suppose  $(\sigma, \gamma) \in P_R$ . Let  $Q$  be the relation modeling  $\text{Lk}(\Psi_R, \sigma)$ .*

*If  $X \notin \Psi_R$ , then the length of a longest informative attribute release sequence for localizing  $\sigma$  in  $R$  is  $\ell(P_Q) + 2$ . If  $X \in \Psi_R$  and  $\sigma \neq X$ , then that length is  $\ell(P_Q) + 1$ .*

*(Note: If  $\sigma = X \in \Psi_R$ , then the length is 0; one can identify  $X$  in  $R$  without observation.)*

Comment: If  $P_Q$  does not contain the top element  $\hat{1}_Q$  of  $P_Q^+$ , then  $\ell(P_Q) + 2 = \ell(P_Q^+)$ , since  $P_Q$  never contains the bottom element  $\hat{0}_Q$ . This occurs precisely when there is no attribute that is shared by all the individuals in the link.

*Proof.* Let us address one special case first, namely when  $\text{Lk}(\Psi_R, \sigma)$  is the empty complex. In that case,  $P_Q$  is empty, so  $\ell(P_Q) = -1$ . Also observe that in this case any  $y \in \gamma$  identifies  $\sigma$ , as otherwise  $\overline{X}$  in the definition of  $Q$  would not be empty. So long as  $\sigma$  is not all of  $X$ , we do indeed have that  $\ell(P_Q) + 2 = 1$ . (Observe, by the way, that it is impossible for the following conditions to be satisfied simultaneously:  $\sigma \subsetneq X \in \Psi_R$  and  $\text{Lk}(\Psi_R, \sigma) = \{\emptyset\}$ .)

Suppose  $\text{Lk}(\Psi_R, \sigma)$  is not the empty complex and that  $X \notin \Psi_R$ . Lemmas 20 and 21 imply that a longest informative attribute release sequence for localizing  $\sigma$  comes from a longest maximal chain in  $P_R^+$  at and above  $(\sigma, \gamma)$  and thus may be obtained by Corollary 78 from a maximal chain in  $P_Q$ . The length of the chain in  $P_Q$  is two shorter than that in  $P_R$ . (Why? Because  $(\sigma, \gamma) \in P_R^+$  becomes  $\hat{0}_Q \in P_Q^+$ , which is not present in  $P_Q$ , and because the top element  $\hat{1}_R = (X, \emptyset) \in P_R^+$  disappears altogether.) So  $\ell(P_Q) + 2$  gives the correct length of the iars in  $R$ .

Suppose  $\text{Lk}(\Psi_R, \sigma)$  is not the empty complex but that  $\sigma \subsetneq X \in \Psi_R$ . The argument proceeds as before except that now the top element of  $P_R^+$  looks like  $\hat{1}_R = (X, \gamma_0)$ , with  $\gamma_0 \neq \emptyset$ . It appears in  $P_R$ . Consequently,  $\hat{1}_Q = (X \setminus \sigma, \gamma_0)$  and  $\hat{1}_Q$  now also appears in  $P_Q$ . So a maximal chain in  $P_Q$  is now only one shorter than a corresponding maximal chain in  $P_R$  at and above  $(\sigma, \gamma)$ , meaning  $\ell(P_Q) + 1$  gives the correct length of a longest iars.  $\square$

### F.3 Isotropy

We turn now to the proof of our isotropy sphere theorem, with the theorem replicated here from earlier in the report. Recall also Definitions 13, 14, 15, 16, and 18 from pages 41–43.

**Theorem 19** (Isotropy = Minimal Identification = Sphere). *Let  $R$  be a relation and suppose  $\emptyset \neq \gamma \in \Phi_R$ . Let  $\sigma = \psi_R(\gamma)$ . Then the following four conditions are equivalent:*

- (a)  $\gamma$  is isotropic.
- (b)  $\gamma$  is minimally identifying (for  $\sigma$ ).
- (c)  $\Psi_{Q(\sigma, \gamma)} \simeq \mathbb{S}^{k-2}$ , with  $k = |\gamma|$ .
- (d)  $\Phi_{Q(\sigma, \gamma)} = \partial(\gamma)$ .

*Proof.* Observe that  $\sigma \in \Psi_R$  and  $\gamma \subseteq (\phi_R \circ \psi_R)(\gamma) = \phi_R(\sigma)$ , so constructing  $Q(\sigma, \gamma)$  is valid.

Note that  $\gamma \notin \Phi_{Q(\sigma, \gamma)}$ . For if there were some  $x \in \overline{X}$  such that  $(x, y) \in Q(\sigma, \gamma) \subseteq R$  for every  $y \in \gamma$ , then  $x \in \sigma$ , but  $\sigma$  is disjoint from  $\overline{X}$ .

If  $|\gamma| = 1$ , then  $\mathbb{S}^{k-2} = \mathbb{S}^{-1} = \{\emptyset\} = \partial(\gamma)$ . Write  $\gamma = \{y\}$ . Then  $\gamma$  is isotropic if and only if  $y$  constitutes an informative attribute release sequence if and only if  $y \notin \phi_R(X)$ . If  $y \in \phi_R(X)$ , then  $\sigma = X$  so our conventions say  $\Psi_{Q(\sigma, \gamma)} = \Phi_{Q(\sigma, \gamma)} = \emptyset \neq \{\emptyset\}$ . Moreover,  $\psi_R(\emptyset) = \sigma$ , so  $\gamma$  is not minimally identifying. If  $y \notin \phi_R(X)$ , then  $\sigma = X_y \subsetneq X$  and  $\overline{X} = \emptyset$ , and both  $\Psi_{Q(\sigma, \gamma)}$  and  $\Phi_{Q(\sigma, \gamma)}$  are instances of  $\{\emptyset\}$ , by our conventions. Moreover,  $\psi_R(\emptyset) \neq \sigma$ . So we see that (a), (b), (c), (d) are all equivalent when  $|\gamma| = 1$ .

Henceforth assume that  $|\gamma| > 1$ . It will be convenient to write  $\gamma = \{y_1, \dots, y_k\}$ , with  $k > 1$ , with the element indexing chosen arbitrarily.

As we have observed elsewhere, (c) and (d) are equivalent by Dowker duality and the fact that only a boundary complex can produce  $\mathbb{S}^{k-2}$  homotopy type when the underlying vertex set has size  $k$ .

We will first show that (a) implies (d) and (b):

Suppose that  $\gamma$  is isotropic.

We wish to show that all proper subsets of  $\gamma$  are simplices in  $\Phi_{Q(\sigma, \gamma)}$ . Without loss of generality, consider  $\{y_1, \dots, y_{k-1}\}$ . If we can show that  $\psi_R(\{y_1, \dots, y_{k-1}\}) \setminus \sigma \neq \emptyset$ , then that provides an  $x \in \overline{X}$  such that  $(x, y_i) \in R$  for  $i = 1, \dots, k-1$ , thereby establishing that  $\{y_1, \dots, y_{k-1}\} \in \Phi_{Q(\sigma, \gamma)}$ . It also establishes that  $\psi_R(\{y_1, \dots, y_{k-1}\}) \supsetneq \sigma$ . Since the “missing element”  $y_k$  is arbitrary in  $\gamma$ , we see that  $\Phi_{Q(\sigma, \gamma)} = \partial(\gamma)$  and that  $\gamma$  is minimally identifying.

Suppose otherwise:  $\psi_R(\{y_1, \dots, y_{k-1}\}) = \sigma = \psi_R(\gamma)$ , so also  $(\phi_R \circ \psi_R)(\{y_1, \dots, y_{k-1}\}) = (\phi_R \circ \psi_R)(\gamma) \supseteq \gamma$ . That says  $y_k \in (\phi_R \circ \psi_R)(\{y_1, \dots, y_{k-1}\})$ , violating the assumption that any ordering of  $\gamma$  is an informative attribute release sequence.

We will now show that (d) implies (a):

Suppose that  $\Phi_{Q(\sigma, \gamma)} = \partial(\gamma)$ .

If some ordering of  $\gamma$  is not an informative attribute release sequence, then we can rearrange the sequence further to establish that the last element is implied by all the others, i.e., that  $y_k \in (\phi_R \circ \psi_R)(\{y_1, \dots, y_{k-1}\})$ . Arguing as we did in the proof of Lemma 20 on page 109, we obtain:

$$\begin{aligned}
 \psi_R(\{y_1, \dots, y_{k-1}\}) &= (\psi_R \circ \phi_R)(\psi_R(\{y_1, \dots, y_{k-1}\})) \\
 &= \psi_R((\phi_R \circ \psi_R)(\{y_1, \dots, y_{k-1}\})) \\
 &= \psi_R(\{y_k\} \cup (\phi_R \circ \psi_R)(\{y_1, \dots, y_{k-1}\})) \\
 &= X_{y_k} \cap \psi_R((\phi_R \circ \psi_R)(\{y_1, \dots, y_{k-1}\})) \\
 &= X_{y_k} \cap \psi_R(\{y_1, \dots, y_{k-1}\}) \\
 &= \psi_R(\{y_1, \dots, y_k\}) \\
 &= \psi_R(\gamma) \\
 &= \sigma.
 \end{aligned}$$

On the other hand, since  $\{y_1, \dots, y_{k-1}\} \in \Phi_{Q(\sigma, \gamma)}$ , there is a witness  $x \in \overline{X}$ , meaning  $x \in \psi_R(\{y_1, \dots, y_{k-1}\})$ , which contradicts  $\overline{X} \cap \sigma = \emptyset$ .

Finally, we will show that (b) implies (d):

Suppose that  $\gamma$  is minimally identifying.

Observe that  $\psi_R(\{y_1, \dots, y_{k-1}\}) \supsetneq \sigma$ . As above, this establishes  $\{y_1, \dots, y_{k-1}\} \in \Phi_{Q(\sigma, \gamma)}$ , from which we conclude that  $\Phi_{Q(\sigma, \gamma)} = \partial(\gamma)$ , since the missing element  $y_k$  was arbitrary.  $\square$

## G Many Long Chains

This appendix provides a proof of Theorem 25 from page 47.

First, we need some tools:

Recall what it means for a poset to be almost a join-based lattice from Definition 24 on page 47.

**Definition 80** (Join Completion). *Suppose  $P$  is almost a join-based lattice. Let  $S$  be a subset of  $P$ . The bounded join-completion of  $S$  in  $P$  is the set  $S^\vee$  defined by:*

$$S^\vee = \{p \in P \mid p \leq s, \text{ some } s \in S, \text{ and } p = s_1 \vee \cdots \vee s_m, \text{ with each } s_i \in S, \text{ and } m \geq 1\}.$$

Here and in the rest of this appendix, “ $\leq$ ” and “ $<$ ” refer to the partial order on  $P$ , and “ $\vee$ ” denotes the join operation on  $P \cup \{\hat{1}\}$ .

We also define  $S_{\max}$  to consist of all the maximal elements of  $S$  relative to the partial order inherited from  $P$ .

The following facts will be useful. Assume  $S \subseteq P$ , with  $P$  almost a join-based lattice. Then:

1.  $S^\vee$  is almost a join-based lattice. The join operation for elements  $p, q \in S^\vee$  is given by:

$$p \vee_{S^\vee} q = \begin{cases} p \vee q, & \text{if } p \vee q \leq s, \text{ for some } s \in S; \\ \hat{1}, & \text{otherwise.} \end{cases}$$

2.  $S \subseteq S^\vee$  and  $S_{\max} = (S^\vee)_{\max}$ .
3.  $(S^\vee)^\vee = S^\vee$ .
4. If  $T \subseteq S$ , then  $T^\vee \subseteq S^\vee$ .
5. If  $T \subseteq S^\vee$  such that  $S_{\max} \setminus T \neq \emptyset$ , then  $T^\vee \subsetneq S^\vee$ .
6. Let  $\emptyset \subsetneq T \subseteq S$ . Then the poset

$$S_T = \{p \in S^\vee \mid p \leq t, \text{ for all } t \in T\}$$

is almost a join-based lattice. The join operation for elements  $p, q \in S_T$  is given by:

$$p \vee_{S_T} q = \begin{cases} p \vee q, & \text{if } p \vee q \leq t, \text{ for all } t \in T; \\ \hat{1}, & \text{otherwise.} \end{cases}$$

7. Fact 6 holds as well for the poset  $S'_T = \{p \in S^\vee \mid p < t, \text{ for all } t \in T\}$ , now using “ $<$ ” in place of “ $\leq$ ” throughout.

**Lemma 81** (Contractibility of Closed Semi-Intervals). *Suppose  $S \subseteq P$ , with  $P$  almost a join-based lattice. Let  $\emptyset \subsetneq T \subseteq S$  and define the poset  $S_T$  as in Fact 6 on page 114.*

*If  $S_T \neq \emptyset$ , then  $S_T$  is contractible.*

*Proof.* Suppose  $p$  and  $q$  are arbitrary elements of  $S_T$ . Every element of  $T$  is an upper bound for both  $p$  and  $q$ . Since  $T$  is not empty, this means  $p \vee q$  exists in  $P$  and  $p \vee q \leq t$  for all  $t \in T$ . Since  $t \in S$ , we have that  $p \vee q \in S^\vee$  and thus  $p \vee q \in S_T$  as well. Consequently, the lattice  $S_T \cup \{\hat{0}, \hat{1}\}$  is noncomplemented, implying that  $S_T$  is contractible, by a fact on page 84.  $\square$

Intuitively:  $\Delta(S_T)$  is a cone with apex  $\bigwedge T$ , the meet of all the upper bounds.

Caution: The lemma need not hold for  $S'_T$  as defined in Fact 7.

We now specialize a topological tool to our current setting. We refer to the lemma as “cycle tightening” because we will apply the lemma with  $p \in S_{\max}$  and with  $z$  a reduced homology generator of  $\Delta(P)$ . The lemma will allow us to move that generator downward in  $P$ .

**Lemma 82** (Cycle Tightening). *Let  $P$  be almost a join-based lattice. Suppose  $z = \sum_i n_i \tau_i$  is a nontrivial reduced  $k$ -cycle for  $\Delta(P)$ , i.e.,  $0 \neq z \in C_k(\Delta(P); \mathbb{Z})$  and  $\tilde{\partial}z = 0$ , for some  $k \geq 0$ .*

*Define  $S = \|z\|$  and  $K = \{\tau \in \Delta(P) \mid \tau \subseteq S^\vee\}$ .*

*Let  $p \in S$ .*

*If  $\tilde{H}_{k-1}(\text{Lk}(K, p); \mathbb{Z}) = 0$ , then there exists  $\eta \in C_{k+1}(\overline{\text{St}}(K, p); \mathbb{Z})$  such that  $p \notin \|z + \tilde{\partial}\eta\|$ , now viewing  $\eta \in C_{k+1}(\Delta(P); \mathbb{Z})$ .*

*Proof.* Let  $W = \overline{\text{St}}(K, p)$  and  $A = \text{Lk}(K, p)$ . Note that  $A$  is not the empty complex (that observation follows from the reduced homology assumption when  $k = 0$  and the fact that  $p$  is part of a simplex containing at least one other element when  $k > 0$ ).

The long exact sequence for a pair [12] therefore gives us the following exact sequence:

$$0 = \tilde{H}_k(W; \mathbb{Z}) \longrightarrow \tilde{H}_k(W, A; \mathbb{Z}) \longrightarrow \tilde{H}_{k-1}(A; \mathbb{Z}) = 0$$

The left 0 comes from  $W$  being a cone and the right 0 comes from the lemma’s hypotheses. Consequently,  $\tilde{H}_k(W, A; \mathbb{Z}) = 0$ . Now let  $z_S$  consist of the part of  $z$  that lies within  $W$ , so:

$$z_S = \sum_{\tau_i \in W} n_i \tau_i.$$

Since  $z$  is a reduced  $k$ -cycle with support in  $\text{verts}(K)$ ,  $z_S$  is a reduced relative  $k$ -cycle for the pair  $(W, A)$ .

Since  $\tilde{H}_k(W, A; \mathbb{Z}) = 0$ ,  $z_S$  must be a reduced relative boundary, so there exists  $\kappa \in C_{k+1}(W; \mathbb{Z})$  such that  $z_S = \tilde{\partial}\kappa + \gamma$ , with  $\gamma \in C_k(A; \mathbb{Z})$ .

Now let  $\eta = -\kappa$  and view  $\eta \in C_{k+1}(\Delta(P); \mathbb{Z})$ .

Observe that  $\|z_S + \tilde{\partial}\eta\| \subseteq \text{verts}(A) \subseteq \text{verts}(\text{dl}(K, p))$ . Consequently,  $p \notin \|z + \tilde{\partial}\eta\|$ .  $\square$

**Lemma 83** (Maximal Element Cardinality). *Let  $P$  be almost a join-based lattice. Suppose  $P$  has reduced integral homology in dimension  $k \geq 0$ , that is,  $\tilde{H}_k(\Delta(P); \mathbb{Z}) \neq 0$ . Consider a reduced homology generator  $z = \sum_i n_i \tau_i$ , for some collection  $\{\tau_i\}$  such that  $n_i \neq 0$  for each  $\tau_i$ .*

*Let  $S = \|z\|$ . Then  $|S_{\max}| \geq k + 2$ .*

*Proof.* Since  $S \subseteq S^\vee$ ,  $z \in C_k(\Delta(S^\vee); \mathbb{Z})$ . If there exists  $\eta \in C_{k+1}(\Delta(S^\vee); \mathbb{Z})$  such that  $\tilde{\partial}\eta = z$ , then  $z$  would also be a reduced boundary in  $\Delta(P)$ . So,  $\tilde{H}_k(\Delta(S^\vee); \mathbb{Z}) \neq 0$  and  $z$  is a reduced homology generator for  $\Delta(S^\vee)$ .

Recall the notation  $S_T$  in Fact 6 on page 114. We claim that

$$\bigcup_{t \in S_{\max}} \Delta(S_{\{t\}}) = \Delta(S^\vee).$$

To see this, first observe that the empty simplex  $\emptyset$  appears in both these sets. Then:

- I. Suppose  $\emptyset \neq \sigma \in \Delta(S_{\{t\}})$  for some  $t \in S_{\max}$ . Being a chain in  $S_{\{t\}}$ , we can write  $\sigma$  as  $\{p_0 < p_1 < \cdots < p_\ell\}$ , for some  $\ell \geq 0$ , with each  $p_i \in S^\vee$  and with  $p_\ell \leq t \in S_{\max} \subseteq S$ .

Consequently,  $\sigma \in \Delta(S^\vee)$  as well.

- II. Suppose  $\emptyset \neq \sigma \in \Delta(S^\vee)$ . Then  $\sigma = \{p_0 < p_1 < \cdots < p_\ell\}$ , for some  $\ell \geq 0$ , with each  $p_i \in S^\vee$ . By definition of  $S^\vee$  and  $S_{\max}$ ,  $p_\ell \leq s \leq t$ , for some  $s \in S$  and  $t \in S_{\max}$ .

Consequently,  $\sigma \in \Delta(S_{\{t\}})$  as well, for that  $t$ .

Similarly, one sees that, for any  $\emptyset \neq T \subseteq S$ ,

$$\bigcap_{t \in T} \Delta(S_{\{t\}}) = \Delta(S_T).$$

The complex on the right is either empty or contractible, by Lemma 81, so we see that the intersection on the left is either empty or contractible.

A variation of the Nerve Lemma now implies that  $\Delta(S^\vee)$  and the nerve of the simplicial complexes  $\{\Delta(S_{\{t\}})\}_{t \in S_{\max}}$  have the same homotopy type (see Theorem 10.6(i) in [1]).

Since  $\Delta(S^\vee)$  has reduced homology in dimension  $k$ , so does the nerve of  $\{\Delta(S_{\{t\}})\}_{t \in S_{\max}}$ .

The nerve of  $\{\Delta(S_{\{t\}})\}_{t \in S_{\max}}$  is isomorphic to a simplicial complex with underlying vertex set  $S_{\max}$ . In order for a simplicial complex to have reduced homology in dimension  $k$  it must have at least  $k + 2$  vertices. Thus  $|S_{\max}| \geq k + 2$ .  $\square$

We now turn to the proof of the main theorem, the statement of which is replicated here:

**Theorem 25** (Many Chains). *Let  $P$  be almost a join-based lattice. Suppose  $P$  has reduced integral homology in dimension  $k \geq 0$ , that is,  $\tilde{H}_k(\Delta(P); \mathbb{Z}) \neq 0$ .*

*Then there are at least  $(k + 2)!$  maximal chains in  $P$  of length at least  $k$ .*

*Proof.* The proof is by induction on  $k$ .

I. For the base case,  $k = 0$ , observe that  $\Delta(P)$  must have at least two vertices that are incomparable in  $P$ , as otherwise  $\Delta(P)$  would be either empty or contractible. Each vertex sits inside a maximal chain of  $P$ . The chains are distinct since the vertices are incomparable.

II. For the induction step, assume that, for some  $k \geq 1$ , the theorem holds for all relevant  $P$  with reduced homology in dimension  $k - 1$ . We need to establish the theorem for all relevant  $P$  with reduced homology in dimension  $k$ .



Let  $z = \sum_i n_i \tau_i$  be a homology generator of  $\tilde{H}_k(\Delta(P); \mathbb{Z})$ , with  $n_i \neq 0$  for all  $\tau_i$ .

Define  $S$  and  $K$  by  $S = \|z\|$  and  $K = \{\tau \in \Delta(P) \mid \tau \subseteq S^\vee\}$ . Interpretation:  $S$  is the support of the homology generator  $z$  and  $K$  is the subcomplex of  $\Delta(P)$  formed by restricting to the bounded join-completion of  $z$ 's support.

We now have an inner induction, which we will describe as an iterative algorithm:  
(Notation: superscript  $(j)$  indicates the  $j^{\text{th}}$  iteration.)

1. Initialize with  $z^{(0)} = z$ ,  $S^{(0)} = S$ , and  $K^{(0)} = K$ .
2. Suppose  $z^{(j)}$ ,  $S^{(j)}$ , and  $K^{(j)}$  have been defined, with  $z^{(j)}$  a homology generator of  $\tilde{H}_k(\Delta(P); \mathbb{Z})$ , and with  $S^{(j)}$  and  $K^{(j)}$  similar in meaning to  $S$  and  $K$ , now based on  $z^{(j)}$ .  
In particular,  $z^{(j)}$  has support  $S^{(j)}$  and all of  $K^{(j)}$ 's vertices lie in  $(S^{(j)})^\vee$ .  
Pick some  $p \in (S^{(j)})_{\max}$  such that  $\tilde{H}_{k-1}(\text{Lk}(K^{(j)}, p); \mathbb{Z}) = 0$ .  
If no such  $p$  exists, then the loop ends.
3. Otherwise, invoke Lemma 82 to find an  $\eta \in C_{k+1}(\overline{\text{St}}(K^{(j)}, p); \mathbb{Z})$  such that  $p \notin \|z^{(j)} + \tilde{\partial}\eta\|$ .  
Let  $z^{(j+1)} = z^{(j)} + \tilde{\partial}\eta$ ,  $S^{(j+1)} = \|z^{(j+1)}\|$ , and

$$K^{(j+1)} = \left\{ \tau \in \Delta(P) \mid \tau \subseteq (S^{(j+1)})^\vee \right\}.$$

Observe that  $S^{(j+1)} \subseteq \|z^{(j)}\| \cup \|\tilde{\partial}\eta\| \subseteq (S^{(j)})^\vee$ .

On the other hand,  $p \in (S^{(j)})_{\max} \setminus S^{(j+1)}$ . So by Fact 5 on page 114,  $(S^{(j+1)})^\vee \subsetneq (S^{(j)})^\vee$ .

In other words, the possible vertex set for the simplicial complex shrinks with each iteration and so the loop must eventually end,  $P$  being finite.

Given this iterative algorithm, we can now assume without loss of generality that  $\tilde{H}_{k-1}(\text{Lk}(K, p); \mathbb{Z}) \neq 0$  for each  $p$  that is a maximal element in the support  $S$  of the given homology generator  $z$ .

Observe that  $\text{Lk}(K, p) = \{\tau \in \Delta(P) \mid \tau \subseteq S^\vee \text{ and } s < p \text{ for every } s \in \tau\}$ , when  $p \in S_{\max}$ .

Consequently,  $\text{Lk}(K, p) = \Delta(Q_p)$ , where  $Q_p$  is the subposet of  $P$  given by

$$Q_p = \{s \in S^\vee \mid s < p\}.$$

By Fact 7 on page 114,  $Q_p$  is itself almost a join-based lattice.

$Q_p$  has reduced integral homology in dimension  $k-1$ , so by the induction hypothesis, there are at least  $(k+1)!$  maximal chains in  $Q_p$  of length at least  $k-1$ . As the description of  $Q_p$  makes clear, we can extend each of these chains in  $P$  by adding  $p$  as a top element, then further refine and/or extend each chain as needed into a maximal chain in  $P$ . Distinct chains remain distinct after this augmentation since the process only adds elements of  $P$  that lie outside  $Q_p$ .

Consequently, we obtain for each  $p \in S_{\max}$  at least  $(k+1)!$  distinct maximal chains in  $P$  of length at least  $k$ , each touching  $p$ . A maximal chain in  $P$  cannot contain more than one element of  $S_{\max}$  since such elements are necessarily incomparable. Letting  $p$  vary over  $S_{\max}$  therefore produces at least  $|S_{\max}| \cdot (k+1)!$  distinct maximal chains in  $P$  of length at least  $k$ .

By Lemma 83,  $|S_{\max}| \geq k+2$ . So  $P$  contains at least  $(k+2)!$  distinct maximal chains of length at least  $k$ .  $\square$

**Corollary 26** (Holes Reduce Inference). *Let  $R$  be a relation. Suppose  $P_R$  has reduced integral homology in dimension  $k \geq 0$ . Then there are at least  $(k+2)!$  maximal chains in  $P_R$  of length at least  $k$ .*

*Proof.* The assertion follows from Theorem 25, since  $P_R$  is almost a join-based lattice.

(The join operation is exactly that of  $P_R^+$ . In particular, the top element  $\hat{1}_R$  of  $P_R^+$  is not already in  $P_R$ , since  $P_R$  has homology, so we may adjoin that as the upper bound  $\hat{1}$  for  $P_R$ .)  $\square$

Recall informative attribute release sequences from Appendix F.

**Corollary 27** (Holes Defer Recognition). *Let  $R$  be a relation and let  $(\sigma, \gamma) \in P_R$ .*

*Define  $Q = Q(\sigma, \gamma)$  as per Definition 18 and recall Definition 16, from pages 42–43.*

*Suppose  $P_Q$  has reduced integral homology in dimension  $k \geq 0$ .*

*Then there are at least  $(k+2)!$  distinct informative attribute release sequences  $y_1, \dots, y_\ell$  for  $R$ , each with  $\ell \geq k+2$ , such that  $\psi_R(\{y_1, \dots, y_\ell\}) = \sigma$ . Consequently,  $r_{\text{slow}}(\sigma) \geq k+2$ .*

*Proof.* By Corollary 26,  $P_Q$  contains at least  $(k+2)!$  maximal chains of length at least  $k$ .

The rest of the argument is much like that in the proof of Corollary 79 on page 111:

- Each maximal chain in  $P_Q$  gives rise to a maximal chain in  $P_R^+$  at or above  $(\sigma, \gamma)$ .
- Distinctness in  $P_Q$  carries over to  $P_R^+$ .
- In moving from  $P_Q$  to  $P_R^+$  one adds two elements:
  1. One adds  $(\sigma, \gamma)$ , corresponding to  $\hat{0}_Q$  in  $P_Q^+$ .
  2.  $P_Q$  has homology, so no attribute is shared by all individuals, either in  $Q$  or  $R$ . One thus also adds the top element  $\hat{1}_R$  of  $P_R^+$ , corresponding to  $\hat{1}_Q$  in  $P_Q^+$ .

Summary: Each distinct maximal chain of  $P_Q$  gives rise to a distinct maximal chain at or above  $(\sigma, \gamma)$  in  $P_R^+$  of length at least  $k+2$ , and therefore a distinct informative attribute release sequence of length at least  $k+2$ . So by Definition 16,  $r_{\text{slow}}(\sigma) \geq k+2$ .

(How do we know that distinct maximal chains produce distinct iars? Because if two iars are the same, the chains must be the same, by the “Moreover” of Lemma 20 on page 109. It is true that one may be able to obtain different iars from the same maximal chain, but our counting was over maximal chains, so provides a lower bound for the number of distinct iars.)  $\square$

**Comment:** Since  $P_Q$  has reduced homology in nonnegative dimension,  $\sigma \neq X$ . Along with the assumption  $(\sigma, \gamma) \in P_R$ , that means relation  $Q(\sigma, \gamma)$  models the link  $\text{Lk}(\Psi_R, \sigma)$ .

## H Obfuscating Strategies

Recall the discussion and terminology of Section 13.

The primary goal of this appendix is to provide a proof of Theorem 31. In addition, this appendix provides proof of some of the assertions in the bullets on pages 68–69.

Once again, we first need to develop some tools:

### H.1 Source Complex

Subsection 13.1 introduced the strategy complex  $\Delta_G$  of a graph  $G = (V, \mathfrak{A})$ . Recall that every action  $a \in \mathfrak{A}$  has a unique source state in  $V$ . Given a set of actions  $\mathcal{A} \subseteq \mathfrak{A}$ , we say  $\text{src}(\mathcal{A})$  is the *start region* of  $\mathcal{A}$ , defined by

$$\text{src}(\mathcal{A}) = \{v \in V \mid v \text{ is the source of some } a \in \mathcal{A}\}.$$

One obtains another simplicial complex from  $G$  via  $\text{src}$ , now on underlying vertex set  $V$ :

$$\overline{\Delta}_G = \{\text{src}(\sigma) \mid \sigma \in \Delta_G\}.$$

We refer to this complex as  $G$ 's *source complex*.

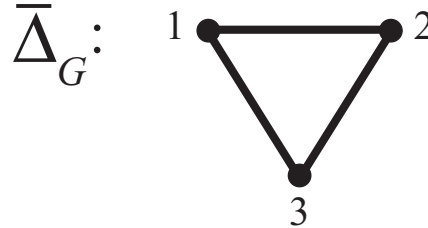


Figure 55: Source complex for the graph of Figure 44 on page 65.

The map  $\text{src} : \mathfrak{F}(\Delta_G) \rightarrow \mathfrak{F}(\overline{\Delta}_G)$  is a homotopy equivalence, so  $\Delta_G \simeq \overline{\Delta}_G$  [6, 7]. Consequently, for a fully controllable graph,  $\overline{\Delta}_G = \partial(V)$ , the boundary complex of the full simplex on vertex set  $V$ . For the graph of Figure 44 on page 65, the source complex is the boundary of a triangle, as shown in Figure 55.

$B$	1	2	3	Goal
$\sigma_1$		•	•	1
$\sigma_2$	•		•	2
$\sigma_3$	•	•		3
$\sigma_4$	•	•		3

Figure 56: Relation  $B$  describes the source complex  $\overline{\Delta}_G$  of the graph of Figure 44. Each row describes the start region of a maximal simplex of  $\Delta_G$ , which appeared in Figure 45 on page 66. The rightmost column again shows each maximal strategy's goal.

In Lemma 29 we saw that  $\Delta_G = \Phi_A$  for the action relation  $A$  defined there. We can see that  $\overline{\Delta}_G = \Phi_B$ , for yet another relation, which we will refer to as  $B$ . Figure 56 shows that relation for the graph of Figure 44. More generally, we have the following lemma:

**Lemma 84.** *Let  $G = (V, \mathfrak{A})$  be a graph as discussed in Section 13 and let  $\mathfrak{M}$  the set of maximal simplices of  $\Delta_G$ . Define relation  $B$  on  $\mathfrak{M} \times V$  by  $B = \{(\sigma, v) \mid v \in \text{src}(\sigma) \text{ and } \sigma \in \mathfrak{M}\}$ .*

*Then  $\Phi_B = \overline{\Delta}_G$ .*

(Again, the proof is nearly definitional, so we omit it.)

(The “ $B$ ” stands for “Beginning” — while “ $S$ ” for “source” might be desirable, we have already used  $S$  to mean “support” elsewhere.)

How should we interpret the remaining Dowker complexes,  $\Psi_A$  and  $\Psi_B$ , for relations  $A$  and  $B$ ? To answer this, let’s look at the semantics of simplices in these complexes. A simplex in  $\Psi_A$  represents a *collection* of maximal simplices of  $\Delta_G$ , namely maximal simplices that have at least one action in common. A simplex in  $\Psi_B$  again represents a collection of maximal simplices of  $\Delta_G$ , now with at least one source state in common. Thus  $\Psi_A \subseteq \Psi_B$ . Moreover, from Dowker duality one obtains:

**Lemma 85.** *Let  $G = (V, \mathfrak{A})$  be a graph as discussed in Section 13, with  $V \neq \emptyset$ .*

*Then the inclusion  $\iota : \mathfrak{F}(\Psi_A) \rightarrow \mathfrak{F}(\Psi_B)$  is a homotopy equivalence.*

Comment: The assumption  $V \neq \emptyset$  means  $\Delta_G$  and  $\overline{\Delta}_G$  are not void, so relation  $B$  is not void. If  $V \neq \emptyset$  but  $\mathfrak{A} = \emptyset$ , then technically relation  $A$  is void, but is is convenient to think of it as an instance of the empty relation instead, with associated empty Dowker complexes.

*Proof.* Consider the following diagram:

$$\begin{array}{ccc}
 \mathfrak{F}(\Psi_A) & \xhookrightarrow{\iota} & \mathfrak{F}(\Psi_B) \\
 \psi_A \uparrow & & \psi_B \uparrow \\
 \mathfrak{F}(\Phi_A) & & \mathfrak{F}(\Phi_B) \\
 \parallel & & \parallel \\
 \mathfrak{F}(\Delta_G) & \xrightarrow{\text{src}} & \mathfrak{F}(\overline{\Delta}_G).
 \end{array}$$

Recall that  $\psi_A$ ,  $\psi_B$ , and  $\text{src}$  are homotopy equivalences.

Let  $\mathfrak{M}$  denote the maximal simplices of  $\Delta_G$ . Observe the following, for each  $\sigma \in \mathfrak{F}(\Delta_G)$ :

$$(\iota \circ \psi_A)(\sigma) = \{\sigma' \in \mathfrak{M} \mid \sigma \subseteq \sigma'\}.$$

$$(\psi_B \circ \text{src})(\sigma) = \{\sigma' \in \mathfrak{M} \mid \text{src}(\sigma) \subseteq \text{src}(\sigma')\}.$$

$$\text{If } \sigma \subseteq \sigma', \text{ then } \text{src}(\sigma) \subseteq \text{src}(\sigma').$$

Consequently,  $(\iota \circ \psi_A)(\sigma) \leq (\psi_B \circ \text{src})(\sigma)$  for every  $\sigma \in \mathfrak{F}(\Delta_G)$ , where “ $\leq$ ” refers to the partial order on  $\mathfrak{F}(\Psi_B)$ .

We conclude that the two order-reversing poset maps  $\iota \circ \psi_A$  and  $\psi_B \circ \text{src}$  are homotopic (see [1], Theorem 10.11) and therefore that  $\iota$  is a homotopy equivalence.  $\square$

**Lemma 86.** *Let  $G = (V, \mathfrak{A})$  be a graph as discussed in Section 13, with  $V \neq \emptyset$ . Then  $\text{src}$  induces a homotopy equivalence of posets  $P_A \rightarrow P_B$  with explicit formula*

$$(\tau, \sigma) \mapsto ((\psi_B \circ \text{src})(\sigma), (\phi_B \circ \psi_B \circ \text{src})(\sigma)).$$

*Proof.* Let  $\text{cl}_A$  denote the image of the closure operator  $\phi_A \circ \psi_A : \mathfrak{F}(\Phi_A) \rightarrow \mathfrak{F}(\Phi_A)$  and let  $\text{cl}_B$  denote the image of the closure operator  $\phi_B \circ \psi_B : \mathfrak{F}(\Phi_B) \rightarrow \mathfrak{F}(\Phi_B)$ . We then have the following diagram of homotopy equivalences:

$$P_A \xrightarrow{\pi_2} \text{cl}_A \xrightarrow{\iota} \mathfrak{F}(\Phi_A) = \mathfrak{F}(\Delta_G) \xrightarrow{\text{src}} \mathfrak{F}(\overline{\Delta}_G) = \mathfrak{F}(\Phi_B) \xrightarrow{\phi_B \circ \psi_B} \text{cl}_B \xrightarrow{\iota} P_B.$$

(Here  $\pi_2$  is projection onto the second coordinate, i.e.,  $\pi_2(\tau, \sigma) = \sigma$  and each of the occurrences of  $\iota$  is an inclusion.)

The composition of all these maps is an order-preserving poset map with the specified formula. The overall map is a homotopy equivalence because each of its constituent maps is a homotopy equivalence.  $\square$

**Corollary 87.** *If  $G$  is fully controllable in Lemma 86, then the formula for the poset map becomes  $(\tau, \sigma) \mapsto ((\psi_B \circ \text{src})(\sigma), \text{src}(\sigma))$ .*

*Proof.* Since  $G$  is fully controllable,  $\Phi_B = \overline{\Delta}_G = \partial(V) \simeq \mathbb{S}^{n-2}$ , with  $n = |V|$ . So  $\Phi_B$  has no free faces, implying that  $\phi_B \circ \psi_B$  is the identity, by Lemma 61 on page 94.  $\square$

**Two Observations:** Assume that  $G$  is a fully controllable graph. (i) No action can appear in all maximal simplices of  $\Delta_G$  as that would mean  $\Delta_G$  would be a cone, so not homotopic to a sphere. Consequently,  $\hat{1}_A = (\mathfrak{M}, \gamma)$  has  $\gamma = \emptyset$  (recall that  $\mathfrak{M}$  is the collection of all maximal simplices of  $\Delta_G$ ). (ii) Even if all actions of  $\mathfrak{A}$  appear individually as vertices of  $\Delta_G$ ,  $\hat{0}_A = (\tau, \mathfrak{A})$  has  $\tau = \emptyset$ , since  $\text{src}(\mathfrak{A}) = V$  and  $V \notin \partial(V)$ .

These observations mean that  $P_A$  does *not* contain either the top element  $\hat{1}_A$  or the bottom element  $\hat{0}_A$  of  $P_A^+$ , when  $G$  is fully controllable.

## H.2 Delaying Strategy and Goal Recognition

We now turn to the proof of the main theorem, the statement of which is replicated here:

**Theorem 31** (Delaying Strategy Identification). *Let  $G = (V, \mathfrak{A})$  be a fully controllable graph, with  $n = |V| > 1$ . Let  $A$  be the relation constructed as in Lemma 29 on page 67 and let  $P_A$  be its associated doubly-labeled poset. Then:*

*For each  $v \in V$ , there exists a maximal strategy  $\sigma_v \in \Delta_G$  for attaining singleton goal state  $v$  such that  $P_A$  contains at least  $(n-1)!$  distinct maximal chains for identifying  $\sigma_v$ , with each chain consisting of at least  $n-1$  elements.*

*Proof.* Let  $P_A^{\text{op}}$  be  $P_A$  but with the opposite order. Then  $P_A^{\text{op}}$  is almost a join-based lattice, with join operation for elements of  $P_A^{\text{op}}$  given by

$$(\tau_1, \sigma_1) \vee (\tau_2, \sigma_2) = \begin{cases} (\tau_1 \cap \tau_2, (\phi_A \circ \psi_A)(\sigma_1 \cup \sigma_2)), & \text{when } \tau_1 \cap \tau_2 \neq \emptyset; \\ \hat{1}, & \text{otherwise.} \end{cases}$$

The maximal elements of  $P_A^{\text{op}}$  are of the form  $(\{\sigma\}, \sigma)$ , with  $\sigma$  varying over the maximal simplices of  $\Delta_G$ . Each minimal element of  $P_A^{\text{op}}$  is of the form  $(\psi_A(\{\mathbf{a}\}), (\phi_A \circ \psi_A)(\{\mathbf{a}\}))$ , with action  $\mathbf{a}$  some vertex of  $\Delta_G$ . (Aside: not every element of that form is necessarily minimal.)

Since  $G$  is fully controllable,  $\Delta(P_A^{\text{op}}) \simeq \mathbb{S}^{n-2}$ , which has reduced homology in dimension  $k = n - 2$ . By the proof of Theorem 25, on page 117, there exists a homology generator  $z$  for  $\Delta(P_A^{\text{op}})$  with support  $S = \|z\|$  such that  $P_A^{\text{op}}$  contains, for each  $p \in S_{\max}$ , a collection of maximal chains passing through  $p$  with the following property: Even if one merely considers the portions of the chains at and below  $p$ , the collection contains at least  $(n - 1)!$  distinct such subchains and each subchain has length at least  $n - 2$ . Each full chain, being maximal, must be a path in  $P_A^{\text{op}}$  between some top element  $(\{\sigma\}, \sigma)$  and some bottom element  $(\psi_A(\{\mathbf{a}\}), (\phi_A \circ \psi_A)(\{\mathbf{a}\}))$ . Working upward from the bottom in  $(P_A^+)^{\text{op}}$  (which is equivalent to working downward from the top in  $P_A^+$ ), each such chain therefore gives rise to an informative action release sequence for identifying  $\sigma$ , consisting of at least  $n - 1$  actions. Moreover, there are at least  $(n - 1)!$  different such sequences for that same strategy  $\sigma$ ; we can hold fixed the portion of any chain at and above  $p$  in  $P_A^{\text{op}}$ , while varying the portion below  $p$  in at least  $(n - 1)!$  different ways.

Let  $p \in S_{\max}$  and suppose  $c$  is some maximal chain of  $P_A^{\text{op}}$  that passes through  $p$  and touches top element  $(\{\sigma\}, \sigma)$ . Pick  $q \in S$ , with  $q \leq p$  (here “ $\leq$ ” is the order on  $P_A^{\text{op}}$ ). Write  $p = (\tau_p, \sigma_p)$  and  $q = (\tau_q, \sigma_q)$ . Even though  $q$  may not be part of chain  $c$ , we can still conclude that  $\sigma_q \subseteq \sigma_p \subseteq \sigma$ . If additionally  $\text{src}(\sigma_q) = V \setminus \{v\}$ , then  $\sigma$  at the top must be a maximal strategy for attaining singleton goal state  $v$ . In order to prove the theorem, it is therefore enough to show that for any  $v \in V$  some such  $q$  exists.

Recall the source relation  $B$  from Lemma 84. Let  $P_B^{\text{op}}$  be  $P_B$  but with the opposite order. Referring back to the notation in the proof of Lemma 86, and using the fact that  $G$  is fully controllable, one sees that  $\Delta(P_B^{\text{op}}) \cong \Delta(\text{cl}_B) = \Delta(\mathfrak{F}(\Phi_B)) = \text{sd}(\partial(V))$ , with “ $\cong$ ” meaning “isomorphic” and “sd” meaning “first barycentric subdivision”. The isomorphism holds by definition of  $P_B$ . The first equality holds because  $\phi_B \circ \psi_B$  is the identity when  $G$  is fully controllable, as we saw in the proof of Corollary 87. The second equality amounts to the definition of first barycentric subdivision, bearing in mind that  $\Phi_B = \overline{\Delta}_G = \partial(V)$ .

The homotopy equivalence of Lemma 86 carries over to this setting as  $\theta : \Delta(P_A^{\text{op}}) \rightarrow \text{sd}(\partial(V))$  and Corollary 87 provides an explicit formula. Specifically, for vertices  $(\tau, \sigma)$  of  $\Delta(P_A^{\text{op}})$ , one has  $\theta(\tau, \sigma) = \text{src}(\sigma)$ .

Since  $\theta$  is a homotopy equivalence, the induced map  $\theta_*$  on reduced homology must map the homology generator  $z$  to a homology generator for the triangulated  $(n - 2)$ -sphere  $\text{sd}(\partial(V))$ . Consequently,  $\|\theta_*(z)\|$  must consist of all nonempty proper subsets of  $V$ . In particular, for each  $v \in V$ , there is some  $q = (\tau_q, \sigma_q) \in \|z\|$  such that  $\text{src}(\sigma_q) = \theta(q) = V \setminus \{v\}$ , as desired.  $\square$

### H.3 Hamiltonian Flexibility

The next lemma establishes the second bullet in the comments on page 68.

**Definition 88** (Complete Strategy). *Let  $G = (V, \mathfrak{A})$  be a graph as discussed in Section 13. A complete strategy for attaining state  $v$  is a strategy  $\sigma$  that has at least one action at every state other than  $v$ . So  $\sigma \in \Delta_G$  and  $\text{src}(\sigma) = V \setminus \{v\}$ .*

**Lemma 89** (Delaying Goal Identification). *Let  $G = (V, \mathfrak{A})$  be a fully controllable graph. Let  $n = |V|$ . Suppose  $s \in V$  is some desired goal state.*

*There exists a sequence of actions  $a_1, a_2, \dots, a_{n-1}$  in  $\mathfrak{A}$  satisfying the following conditions:*

- (i)  *$\{a_1, \dots, a_{n-1}\}$  is a complete strategy for attaining  $s$ .*
- (ii) *For each  $i = 1, \dots, n-1$ , let  $\sigma_i = \{a_1, \dots, a_i\}$  and  $W_i = \text{src}(\sigma_i)$ . Then for each  $v \in V \setminus W_i$ , there exists a complete strategy  $\sigma$  for attaining  $v$ , such that  $\sigma_i \subseteq \sigma \in \Delta_G$ .*

**Comments:**

(a) Condition (i) implies that no two of the actions  $a_1, \dots, a_{n-1}$  have the same source state.

(b) Condition (ii) further implies that the sequence  $a_1, \dots, a_{n-1}$  forms an informative attribute release sequence for the relation  $A$  defined in Lemma 29 on page 67. The reason is that any state in  $v \in V \setminus W_i$  could still be a goal state after an observer has seen the actions  $a_1, \dots, a_i$ , so the observer cannot predict even the source of the next action to be released.

*Proof.* For the proof, we assume that  $\mathfrak{A}$  contains only deterministic and nondeterministic actions, not stochastic ones. The proof generalizes to graphs that include stochastic actions (in addition to deterministic and nondeterministic) by an argument in [7]. The essence of that argument is that the source complex of a graph does not change if one replaces stochastic transitions by deterministic ones.

We sketch the rest of the proof, assuming all actions are deterministic or nondeterministic.

Since  $G$  is fully controllable, for each state in  $V$  there must be a deterministic transition to that state (from some other state). Backchaining such transitions gives rise to a cycle of deterministic actions, since the graph is finite. If that cycle is Hamiltonian, then we may chose  $a_1, \dots, a_{n-1}$  to be any ordering of those  $n$  deterministic actions except that we omit the action whose source is  $s$ .

Suppose instead that the cycle of deterministic actions covers only a proper subset  $W$  of the state space  $V$ . Form a quotient graph with state space  $V' = \{\diamond\} \cup V \setminus W$ , where  $\diamond$  represents all of  $W$  collapsed to a point. Inductively, the lemma's assertions hold for the quotient graph. One then needs to show how to combine the actions determined by the quotient graph with the cycle on  $W$  in order to satisfy the lemma's assertions for the original graph  $G$ . That argument is straightforward if a bit tedious, so we omit it.  $\square$

The next lemma establishes the Hamiltonian “yes” in the first bullet on page 68.

**Definition 90** (Hamiltonian Action Cycle). *Let  $G = (V, \mathfrak{A})$  be a graph whose actions may have uncertain outcomes. A sequence of actions  $a_1, \dots, a_n$ , with  $n = |V|$ , is a Hamiltonian cycle of actions whenever:*

- (i) *No two actions have the same source state.*
- (ii) *Each action is either deterministic or stochastic (so, nondeterministic is disallowed).*
- (iii) *The source of action  $a_{i+1}$  is a target of action  $a_i$ , for all  $i = 1, \dots, n-1$ , and  $\text{src}(a_1)$  is a target of  $a_n$ .*

**Observe:** Any proper subset of a Hamiltonian cycle of actions is a simplex in  $\Delta_G$ .

(That observation requires understanding the definition of  $\Delta_G$  when stochastic actions are involved: stochastic cycles are fine, so long as they are not recurrent. See [7] for details.)

**Lemma 91** (Delaying Identification of a Given Strategy). *Let  $G = (V, \mathfrak{A})$  be a fully controllable graph. Assume  $\mathfrak{A}$  contains a Hamiltonian cycle of actions  $a_1, \dots, a_n$ , with  $n = |V|$ .*

*Suppose  $\sigma_v$  is a maximal and complete strategy in  $\Delta_G$  for attaining  $v \in V$ . Then  $\sigma_v$  contains actions  $b_1, \dots, b_{n-1}$  that constitute a complete strategy for attaining  $v$  and that form an informative attribute release sequence for relation  $A$ .*

*(Recall: Relation  $A$  was defined in Lemma 29 on page 67; it models the maximal simplices of  $\Delta_G$  in terms of their constituent actions.)*

*Proof.* Let  $\sigma_v$  be as specified.

We can assume without loss of generality that  $V = \{1, \dots, n\}$ , that  $\text{src}(a_i) = i$  for all  $i \in V$ , and that  $v = n > 1$ .

Now let  $b_1, \dots, b_{n-1}$  be any actions in  $\sigma_v$  chosen so that  $\text{src}(b_i) = i$ , for  $i = 1, \dots, n-1$ . (If  $b_i = a_i$  for some  $i$ , that is fine.)

Then  $\{b_1, \dots, b_{n-1}\}$  is itself a complete strategy for attaining  $v$ .

We claim that the release order  $b_{n-1}, \dots, b_1$  constitutes an informative attribute release sequence for relation  $A$ . In fact, we will prove the stronger assertion:

**Claim:** Pick some  $i \in \{1, \dots, n\}$ . Then:

For each  $s \in \{n\} \cup \{1, \dots, i-1\}$ , there exists a complete strategy

$\sigma_s \in \Delta_G$  for attaining  $s$ , with  $\{b_i, b_{i+1}, \dots, b_{n-1}\} \subseteq \sigma_s$ .

(Notation:  $\{b_i, b_{i+1}, \dots, b_{n-1}\} = \emptyset$  when  $i = n$ .)

Consequently, after an observer has seen  $b_{n-1}, \dots, b_i$ , the observer cannot predict even the source state of the next action to be released, and so the action sequence is informative for  $A$ .

The claim certainly holds for  $s = n$ , using the original  $\sigma_v$ . Now consider an  $s \in \{1, \dots, i-1\}$  and let  $\sigma_s = \{a_1, \dots, a_{s-1}\} \cup \{b_{s+1}, \dots, b_{n-1}\} \cup \{a_n\}$ . By arguments from [7],  $\sigma_s \in \Delta_G$ .  $\square$

**Caution:** As mentioned on page 68, just because  $b_{n-1}, \dots, b_1$  as produced by Lemma 91 is an informative attribute release sequence for  $A$ , that does not mean one should always release actions in that fashion. If the release protocol were so rigid, an adversary familiar with the protocol would be able to infer much about the goal. In particular, the target set of  $b_{n-1}$  includes the goal state, so if that action is deterministic, then the adversary would be able to infer the goal from the first action released.



## I Morphisms and Lattice Generators

The aim of this appendix is to prove the claims of Section 14, ending with Theorem 40. That theorem shows how a surjective morphism of relations can use lattice operations to fully cover the image lattice even when the poset map induced by the morphism is not itself surjective.

### I.1 Morphisms

Notation reminder: We frequently will be working with two relations:  $R$  is a relation on  $X^R \times Y^R$  and  $Q$  is a relation on  $X^Q \times Y^Q$ . In order to distinguish rows and columns between the two, we also use notation of the form  $X_y^R$ ,  $Y_x^R$ ,  $X_y^Q$ , and  $Y_x^Q$ .

Now recall the definition of *morphism* from page 70:

**Definition 32** (Morphism). *Let  $R$  be a relation on  $X^R \times Y^R$  and let  $Q$  be a relation on  $X^Q \times Y^Q$ . A morphism of relations  $f : R \rightarrow Q$  is a pair of set functions:*

$$\begin{aligned} f_X &: X^R \rightarrow X^Q \\ f_Y &: Y^R \rightarrow Y^Q \end{aligned}$$

*such that  $(f_X(x), f_Y(y)) \in Q$  whenever  $(x, y) \in R$ .*

Throughout this appendix, 'morphism' refers to Definition 32. When the time comes, we will refer to 'G-morphism' explicitly (see again Definitions 35 and 38 on pages 75 and 76).

**Morphism Equality:** Before proving properties about morphisms, we should give a notion of morphism equality. Suppose  $g, h : R \rightarrow Q$  are two morphisms. We will say that  $g = h$  if and only if  $(g_X(x), g_Y(y)) = (h_X(x), h_Y(y))$  for all  $(x, y) \in R$ . In particular, we do not care what the constituent set maps do on elements that are not relevant to the relations viewed as sets of pairs. (Note: The condition stated is equivalent to requiring  $g_X(x) = h_X(x)$  and  $g_Y(y) = h_Y(y)$  for all  $(x, y) \in R$ .)

The following lemma shows that the component maps of a morphism between relations may be viewed as simplicial maps:

**Lemma 33** (Induced Simplicial Maps). *A morphism  $f : R \rightarrow Q$  between nonvoid relations induces simplicial maps between the Dowker complexes:*

$$\begin{aligned} f_X &: \Psi_R \rightarrow \Psi_Q \\ f_Y &: \Phi_R \rightarrow \Phi_Q \end{aligned}$$

*Proof.* We need to show that  $f_X(\sigma) \in \Psi_Q$  for all  $\sigma \in \Psi_R$ .

If  $\sigma = \emptyset$ , then  $f_X(\sigma) = \emptyset \in \Psi_Q$  since  $Q$  is nonvoid.

If  $\sigma = \{x_1, \dots, x_k\}$ , then  $f_X(\sigma) = \{f_X(x_1), \dots, f_X(x_k)\}$ .

Since  $\emptyset \neq \sigma \in \Psi_R$ , there exists  $y \in Y^R$  such that  $(x, y) \in R$  for all  $x \in \sigma$ . Thus  $(f_X(x), f_Y(y)) \in Q$  for all  $x \in \sigma$ , by the definition of morphism. So  $f_Y(y)$  is a witness for  $f_X(\sigma)$  in  $Q$ , telling us  $f_X(\sigma) \in \Psi_Q$ .

The argument for the map  $f_Y : \Phi_R \rightarrow \Phi_Q$  is similar. □

**Comment:** The nonvoid requirement is an artifact, arising because we sometimes regard void relations as having empty rather than void Dowker complexes, in the context of links (see Definition 7 on page 27, Definition 18 on page 43, the comments about void relations on page 85, and the hypotheses of Lemma 54 on page 90). The nonvoid requirement of Lemma 33 avoids having to worry about mapping from an artificially empty complex into a void one.

**Lemma 34** (Morphism Properties). *Assume the notation from above and that all relevant relations are nonvoid. Let  $f : R \rightarrow Q$  be a morphism of relations. Then:*

(i)  $f_X$  and  $f_Y$  are one-to-one set maps  $\implies f$  is one-to-one  $\iff f$  is a monomorphism.

(ii)  $f$  surjective  $\implies f$  epimorphism  $\iff f_X$  and  $f_Y$  are surjective set maps.

(Additional conditions for that last  $\iff$  : The  $\implies$  direction assumes that  $Q$  has no blank rows or columns, while the  $\impliedby$  direction assumes that  $R$  has no blank rows or columns.)

The two uni-directional implications  $\implies$  above are strict.

(iii) If  $f_X : \Psi_R \rightarrow \Psi_Q$  is surjective and  $Q$  has no blank rows, then  $f_X : X^R \rightarrow X^Q$  is surjective.

Similarly for  $f_Y$ , now assuming that  $Q$  has no blank columns.

The converses need not hold. Indeed,  $f$  itself can be surjective but the maps of simplicial complexes need not be (as we saw with the maps of page 73 and as one can see with simpler examples as well).

(iv) If  $f_X : X^R \rightarrow X^Q$  is one-to-one, then  $f_X : \Psi_R \rightarrow \Psi_Q$  is injective. The converse holds if  $R$  has no blank rows.

Similarly for  $f_Y$ , now assuming that  $R$  has no blank columns for the converse.

*Proof.* We will prove the various implications. Strictness, i.e., failure of converses, where mentioned above, can be seen readily with simple examples.

Part (i):

(a) Let  $f_X$  and  $f_Y$  be one-to-one set maps.

Suppose  $(f_X(x'), f_Y(y')) = (f_X(x), f_Y(y))$ . Then  $f_X(x') = f_X(x)$ , so  $x' = x$ .

And  $f_Y(y') = f_Y(y)$ , so  $y' = y$ . So  $f$  is one-to-one as a set map of pairs.

(b) Let  $f$  be one-to-one as a set map of pairs.

Suppose  $g, h : S \rightarrow R$  are morphisms such that  $f \circ g = f \circ h$ .

Suppose  $(x, y) \in S$ . By assumption,  $(f_X(g_X(x)), f_Y(g_Y(y))) = (f_X(h_X(x)), f_Y(h_Y(y)))$ .

Since  $f$  is one-to-one,  $(g_X(x), g_Y(y)) = (h_X(x), h_Y(y))$ .

So  $g = h$ , by our notion of equality, meaning  $f$  is a monomorphism.

(c) Let  $f$  be a monomorphism.

Suppose  $f(x, y) = f(x', y')$  but  $(x, y) \neq (x', y')$ . Let  $S$  be the relation consisting of the single element  $\{(I, \alpha)\}$ , with  $I$  and  $\alpha$  new symbols:

$$\begin{array}{c|c} S & \alpha \\ \hline I & \bullet \end{array}$$

Define two morphisms  $g, h : S \rightarrow R$  by:

$$\begin{array}{ll} g_X : I \mapsto x & h_X : I \mapsto x' \\ g_Y : \alpha \mapsto y & h_Y : \alpha \mapsto y' \end{array}$$

Then  $g \neq h$ , but  $f \circ g = f \circ h$ , a contradiction. So  $f$  is one-to-one.

Part (ii):

(a) Let  $f$  be surjective as a set map of pairs.

Suppose  $g, h : Q \rightarrow S$  are morphisms such that  $g \circ f = h \circ f$ .

Suppose  $(x', y') \in Q$ .

By surjectivity, there exists  $(x, y) \in R$  such that  $(f_X(x), f_Y(y)) = (x', y')$ . So:

$$(g_X(x'), g_Y(y')) = (g_X(f_X(x)), g_Y(f_Y(y))) = (h_X(f_X(x)), h_Y(f_Y(y))) = (h_X(x'), h_Y(y')).$$

Thus  $g = h$  and we see that  $f$  is an epimorphism.

(b) Assume  $Q$  has no blank rows or columns and let  $f$  be an epimorphism.

Suppose  $f_Y$  is not surjective, so there exists  $y^* \in Y^Q \setminus (f_Y(Y^R))$ .

Let  $S$  be the relation consisting of two elements  $\{(I, \alpha), (I, \beta)\}$ , with  $I, \alpha, \beta$  new symbols:

$$\begin{array}{c|cc} S & \alpha & \beta \\ \hline I & \bullet & \bullet \end{array}$$

Define two morphisms  $g, h : Q \rightarrow S$  by:

$$\begin{array}{lll} g_X(x) = I & h_X(x) = I & \text{for every } x \in X^Q \\ g_Y(y) = \alpha & h_Y(y) = \alpha & \text{for every } y \in Y^Q \setminus \{y^*\} \\ g_Y(y^*) = \alpha & h_Y(y^*) = \beta & \end{array}$$

Since  $y^* \in Y^Q$  and  $Q$  has no blank columns there is at least one  $x^* \in X^Q$  such that  $(x^*, y^*) \in Q$ . So  $g \neq h$ .

Observe that  $g \circ f = h \circ f$  since  $y^*$  does not appear in the image of  $f_Y$ , contradicting  $f$  being an epimorphism.

The argument showing that  $f_X$  is surjective is similar.

(c) Assume  $R$  has no blank rows or columns and let  $f_X$  and  $f_Y$  be surjective.

Suppose  $g, h : Q \rightarrow S$  are morphisms such that  $g \circ f = h \circ f$ .

Suppose  $(x, y) \in Q$ . We need to show that  $g_X(x) = h_X(x)$  and  $g_Y(y) = h_Y(y)$ , as that means  $g = h$ , given our definition of equality. We will make the argument for the  $X$  coordinate; the  $Y$  argument is similar.

Since  $f_X$  is surjective, there exists  $\bar{x} \in X^R$  such that  $f_X(\bar{x}) = x$ . Since  $R$  has no blank rows, there exists  $\bar{y} \in Y^R$  such that  $(\bar{x}, \bar{y}) \in R$ .

Since  $(g \circ f)(\bar{x}, \bar{y}) = (h \circ f)(\bar{x}, \bar{y})$ , one obtains  $g_X(x) = g_X(f_X(\bar{x})) = h_X(f_X(\bar{x})) = h_X(x)$ .

Part (iii):

Suppose  $Q$  has no blank rows and suppose  $f_X : \Psi_R \rightarrow \Psi_Q$  is surjective as a simplicial map.

Suppose  $x \in X^Q$ . Since  $Q$  has no blank rows,  $x$  is a vertex of  $\Psi_Q$ , so there is some  $\bar{x}$  that is a vertex of  $\Psi_R$  such that  $f_X(\bar{x}) = x$ . That says  $f_X : X^R \rightarrow X^Q$  is also surjective as a set map.

The argument for  $f_Y$  assuming  $Q$  has no blank columns is similar.

Part (iv):

(a) Let  $f_X$  be one-to-one as a set map  $X^R \rightarrow X^Q$ . Consider  $f_X$  as a simplicial map  $\Psi_R \rightarrow \Psi_Q$ .

Suppose  $f_X(\sigma) = \kappa = f_X(\tau)$ , with  $\sigma, \tau \in \Psi_R$  and  $\kappa \in \Psi_Q$ .

If  $\kappa = \emptyset$ , then necessarily  $\sigma = \tau = \emptyset$ . Otherwise,  $\sigma \neq \emptyset$  and  $\tau \neq \emptyset$ , so let  $x \in \sigma$ . Then  $f_X(x) \in \kappa$ . So there exists  $x' \in \tau$  such that  $f_X(x') = f_X(x)$ . Since  $f_X$  is one-to-one as a set map, that says  $x' = x$ . Thus  $\sigma \subseteq \tau$ . A similar argument shows the reverse inclusion, so  $\sigma = \tau$ . Thus  $f_X$  is injective as a simplicial map.

(b) Assume  $R$  has no blank rows and let  $f_X$  be injective as a simplicial map  $\Psi_R \rightarrow \Psi_Q$ .

Consider  $f_X$  as a set map  $X^R \rightarrow X^Q$  and suppose  $f_X(x) = f_X(x')$ . Since  $R$  has no blank rows, both  $x$  and  $x'$  are vertices in  $\Psi_R$ . That means  $f_X(\{x\}) = f_X(\{x'\})$  when we view  $f_X$  as a simplicial map, so  $\{x\} = \{x'\}$  by injectivity, i.e.,  $x = x'$ . So we see that  $f_X$  is one-to-one as a set map.

A similar argument holds for the assertions regarding  $f_Y$ . □

## I.2 G-Morphisms

Recall the material of Section 14.4.

**Lemma 36** (Containment). *Let  $f : R \rightarrow Q$  be a morphism of nonvoid relations. Then:*

$$(a) (f_Y \circ \phi_R)(\sigma) \subseteq (\phi_Q \circ f_X)(\sigma), \text{ for every } \sigma \in \Psi_R,$$

$$(b) (f_X \circ \psi_R)(\gamma) \subseteq (\psi_Q \circ f_Y)(\gamma), \text{ for every } \gamma \in \Phi_R.$$

*Proof.* Observe that  $(f_Y \circ \phi_R)(\emptyset) = f_Y(Y^R) \subseteq Y^Q = \phi_Q(\emptyset) = (\phi_Q \circ f_X)(\emptyset)$ .

Now let  $\emptyset \neq \sigma \in \Psi_R$ . Let  $y \in \phi_R(\sigma) \neq \emptyset$ . Then  $(x, y) \in R$  for every  $x \in \sigma$ . Thus  $(f_X(x), f_Y(y)) \in Q$  for every  $x \in \sigma$ . So  $f_Y(y) \in \phi_Q(f_X(\sigma))$ . This is true for all  $y \in \phi_R(\sigma)$ , telling us  $f_Y(\phi_R(\sigma)) \subseteq \phi_Q(f_X(\sigma))$ .

The argument for assertion (b) is similar. □

**Corollary 37** (Homotopic Face Maps). *Let  $f : R \rightarrow Q$  be a morphism of nonvoid relations. Then:*

$$(a) f_X \text{ and } \psi_Q \circ f_Y \circ \phi_R \text{ are homotopic poset maps } \mathfrak{F}(\Psi_R) \rightarrow \mathfrak{F}(\Psi_Q),$$

$$(b) f_Y \text{ and } \phi_Q \circ f_X \circ \psi_R \text{ are homotopic poset maps } \mathfrak{F}(\Phi_R) \rightarrow \mathfrak{F}(\Phi_Q).$$

*Proof.* Let  $\sigma \in \mathfrak{F}(\Psi_R)$ .

By Lemma 36,  $(f_Y \circ \phi_R)(\sigma) \subseteq (\phi_Q \circ f_X)(\sigma)$ .

Therefore  $(\psi_Q \circ f_Y \circ \phi_R)(\sigma) \supseteq (\psi_Q \circ \phi_Q \circ f_X)(\sigma)$ .

So  $(\psi_Q \circ f_Y \circ \phi_R)$  and  $(\psi_Q \circ \phi_Q \circ f_X)$  are homotopic maps (see [1], Theorem 10.11).

Since  $\psi_Q \circ \phi_Q$  is homotopic to the identity on  $\mathfrak{F}(\Psi_Q)$ , part (a) follows.

The proof of (b) is similar. □

**Corollary 39** (Homotopic Poset Maps). *Let  $f : R \rightarrow Q$  be a morphism of nonvoid relations. The induced  $G$ -morphisms  $f_X^g, f_Y^g : P_R \rightarrow P_Q$  are homotopic. (See Definition 38 on page 76.)*

*Proof.* See Figure 49 on page 75 for the underlying maps comprising the  $G$ -morphisms. The  $G$ -morphisms are defined as follows:

For all  $(\sigma, \gamma) \in P_R$ :

$$f_X^g(\sigma, \gamma) = (\sigma', \gamma'), \quad \text{with } \sigma' = (\psi_Q \circ f_Y \circ \phi_R)(\sigma) \quad \text{and} \quad \gamma' = \phi_Q(\sigma').$$

$$f_Y^g(\sigma, \gamma) = (\sigma'', \gamma''), \quad \text{with } \gamma'' = (\phi_Q \circ f_X \circ \psi_R)(\gamma) \quad \text{and} \quad \sigma'' = \psi_Q(\gamma'').$$

These definitions make sense because  $f_X$  and  $f_Y$  map nonempty simplices to nonempty simplices and because the images of  $\psi_Q$  and  $\phi_Q$  may be viewed as lying in  $P_Q$ , by Corollary 45 on page 87. (Similarly, the images of  $\psi_R$  and  $\phi_R$  may be viewed as lying in  $P_R$  — In fact, as used above, these maps are simply switching between the  $\sigma$  and  $\gamma$  components (“labels”) of the given element  $(\sigma, \gamma)$  in  $P_R$ .) Observe that  $f_X^g$  and  $f_Y^g$  are order-preserving poset maps.

Applying Lemma 36 and since  $(\sigma, \gamma) \in P_R$ :

$$(f_Y \circ \phi_R)(\sigma) \subseteq (\phi_Q \circ f_X)(\sigma) = (\phi_Q \circ f_X \circ \psi_R)(\gamma) = \gamma''.$$

Consequently:

$$\sigma' = (\psi_Q \circ f_Y \circ \phi_R)(\sigma) \supseteq \psi_Q(\gamma'') = \sigma''.$$

So the maps are homotopic (see [1], Theorem 10.11). □

### I.3 Lattice Generators

We turn now to the main result.

(Recall that a relation is *tight* when it has no blank rows or columns.)

**Lemma 92** (Generators in Image). *Let  $f : R \rightarrow Q$  be a surjective morphism between nonvoid tight relations.*

*Suppose  $q \in P_Q$  is of the form  $(X_y^Q, (\phi_Q \circ \psi_Q)(\{y\}))$ , for some  $y \in Y^Q$ .*

*Then there exist  $q_1, \dots, q_k$  in the image of  $f_X^g : P_R \rightarrow P_Q$ , with  $k \geq 1$ , such that  $q = \bigvee_{i=1}^k q_i$ . (Here  $\bigvee$  is the join operation of  $P_Q^+$ .)*

*Proof.* By Lemma 34(iii), the component functions  $f_X : X^R \rightarrow X^Q$  and  $f_Y : Y^R \rightarrow Y^Q$  are surjective. Since  $f_Y$  is surjective,  $f_Y^{-1}(\{y\}) = \{y_1, \dots, y_k\} \subseteq Y^R$ , for some  $k \geq 1$ .

For each  $i = 1, \dots, k$ , observe and define the following:

- Since  $R$  has no blank columns,  $X_{y_i}^R \neq \emptyset$ , so  $(X_{y_i}^R, (\phi_R \circ \psi_R)(\{y_i\})) \in P_R$ .
- Define  $\sigma_i$  as the “ $\sigma'$ -component” of  $f_X^g(X_{y_i}^R, (\phi_R \circ \psi_R)(\{y_i\}))$ , meaning:

$$\sigma_i = (\psi_Q \circ f_Y \circ \phi_R)(X_{y_i}^R) = \psi_Q(\gamma) = \bigcap_{\bar{y} \in \gamma} X_{\bar{y}}^Q, \quad \text{with } \gamma = f_Y((\phi_R \circ \psi_R)(\{y_i\})).$$

- Observe that  $y \in \gamma$  since  $y = f_Y(y_i)$  and  $y_i \in (\phi_R \circ \psi_R)(\{y_i\})$ . Therefore  $\sigma_i \subseteq X_y^Q$ .
- Define  $q_i = (\sigma_i, \gamma_i) \in P_Q$ , with  $\gamma_i = \phi_Q(\sigma_i)$ . So  $q_i$  is in the image of  $f_X^g : P_R \rightarrow P_Q$ .

We need to show that  $q = \bigvee_{i=1}^k q_i$ . Expanding, we see:

$$\bigvee_{i=1}^k q_i = \left( (\psi_Q \circ \phi_Q) \left( \bigcup_{i=1}^k \sigma_i \right), \bigcap_{i=1}^k \gamma_i \right).$$

By a bullet above,  $\bigcup_{i=1}^k \sigma_i \subseteq X_y^Q$ , so:

$$\bigcup_{i=1}^k \sigma_i \subseteq (\psi_Q \circ \phi_Q) \left( \bigcup_{i=1}^k \sigma_i \right) \subseteq (\psi_Q \circ \phi_Q)(X_y^Q) = X_y^Q.$$

We will establish  $X_y^Q \subseteq \bigcup_{i=1}^k \sigma_i$ , thereby completing the proof.

Let  $\bar{x} \in X_y^Q$ .

So  $(\bar{x}, y) \in Q$ . By surjectivity, there exists  $(\hat{x}, \hat{y}) \in R$  such that  $f_X(\hat{x}) = \bar{x}$  and  $f_Y(\hat{y}) = y$ .

Now  $\hat{y} = y_j$ , for some  $j \in \{1, \dots, k\}$ , as defined earlier. Thus  $\hat{x} \in X_{y_j}^R$ .

Consequently, for every  $z \in (\phi_R \circ \psi_R)(\{y_j\})$ ,  $(\hat{x}, z) \in R$  and so  $(f_X(\hat{x}), f_Y(z)) \in Q$ .

That means  $(\bar{x}, \bar{y}) \in Q$  for every  $\bar{y} \in f_Y((\phi_R \circ \psi_R)(\{y_j\}))$ .

Therefore,  $\bar{x} \in \sigma_j \subseteq \bigcup_{i=1}^k \sigma_i$  and we conclude that  $X_y^Q \subseteq \bigcup_{i=1}^k \sigma_i$ .

(Note:  $X_y^Q$  need not lie in a single  $\sigma_j$ , since  $j$  depends on  $\bar{x}$ .) □

**Corollary 93.** Assume the hypotheses of Lemma 92.

Suppose further that for some  $y_i \in f_Y^{-1}(\{y\})$ ,  $(\phi_R \circ \psi_R)(\{y_i\}) = \{y_i\}$ .

Then  $q$  is itself in the image of  $f_X^g : P_R \rightarrow P_Q$ .

*Proof.* In the proof Lemma 92, we see that now  $f_Y((\phi_R \circ \psi_R)(\{y_i\})) = \{y\}$ , so  $\sigma_i = X_y^Q$ . □

**Comment:** Corollary 93 helps to explain the example of pages 73 and 77, in which a surjective morphism generated the entire image poset even though the induced maps on the Dowker complexes were not surjective. Namely:

In the Möbius relation  $M$  of page 72, singletons are unmoved by the closure operators. In the tetrahedral relation  $T$  of page 45, maximal simplices are dual to singletons. Intersections of all the maximal simplices in the tetrahedral relation generate all of  $P_T$ . These maximal simplices come from dualizing images of singletons of the Möbius relation.

More specifically: Even though one merely has  $f_X(\{1, 2, 5\}) = \{1, 4\}$ , it is also true that  $f_X^g(\{1, 2, 5\}, \{\mathbf{a}\}) = (\{1, 3, 4\}, \{\mathbf{a}\})$ . The G-morphism  $f_X^g$  therefore supplies the apparently missing simplex  $\{1, 3, 4\}$ .

More generally, the following theorem describes the process:

**Theorem 40** (Lattice Surjectivity). *Let  $R$  and  $Q$  be tight nonvoid relations. Suppose  $f : R \rightarrow Q$  is a surjective morphism. For any  $q \in P_Q$ :*

$$q = \bigwedge_j \bigvee_i q_{ji}, \quad \text{with each } q_{ji} \text{ in the image of } f_X^g : P_R \rightarrow P_Q,$$

$$q = \bigvee_k \bigwedge_\ell q'_{k\ell}, \quad \text{with each } q'_{k\ell} \text{ in the image of } f_Y^g : P_R \rightarrow P_Q.$$

*Proof.* Write  $q = (\sigma, \gamma) \in P_Q$ . Then  $\sigma = \psi_Q(\gamma) = \bigcap_{y \in \gamma} X_y^Q$ .

So  $q = \bigwedge_{y \in \gamma} q_y$ , with each  $q_y \in P_Q$  of the form  $(X_y^Q, (\phi_Q \circ \psi_Q)(\{y\}))$ .

By Lemma 92, for each  $y \in \gamma$ , we have that  $q_y = \bigvee_i q_{y,i}$  with each  $q_{y,i}$  in the image of  $f_X^g : P_R \rightarrow P_Q$  and with  $i$  in some finite index set  $\mathcal{I}(y)$ . Thus:

$$q = \bigwedge_{y \in \gamma} \bigvee_{i \in \mathcal{I}(y)} q_{y,i}.$$

The other form follows by dualizing the previous arguments. □

## J A Few More Examples

### J.1 Local Spheres versus Global Contractibility

The reader may wonder whether privacy preservation always requires a relation to exhibit homology in its Dowker complexes. The answer is that links of individuals must have homology, by Theorems 9 and 10 on pages 28 and 29, but the overall relation need not.

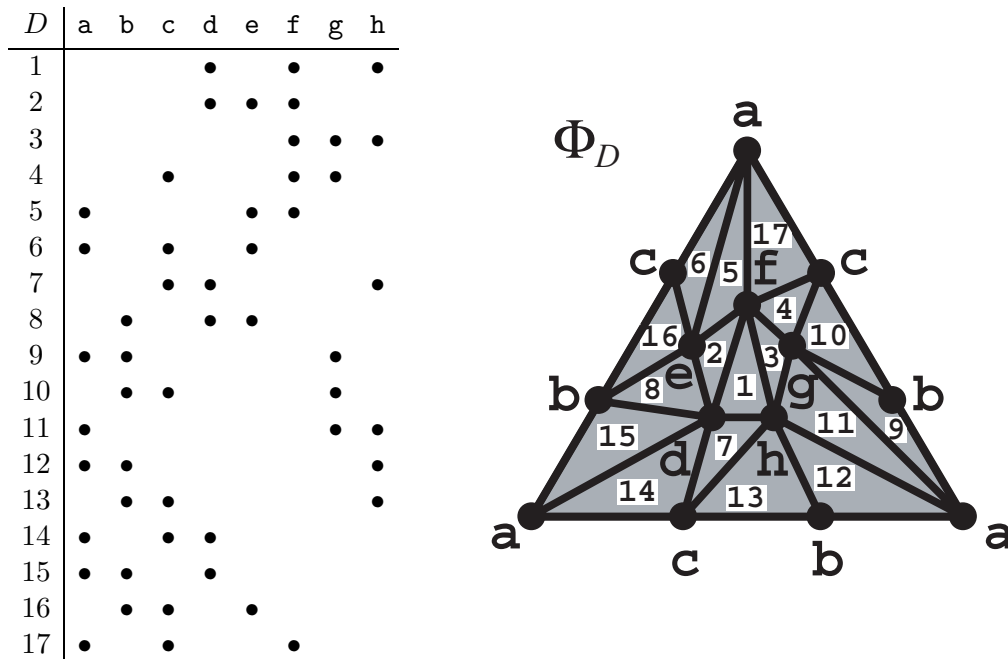


Figure 57: Relation  $D$  and its Dowker complex  $\Phi_D$ . The complex is a triangulation of the Dunces Hat, a contractible space (the seemingly bounding edges actually touch, as suggested by the vertex labels). The Dunces Hat has no free faces, indicating that  $D$  preserves attribute privacy. (Vertices of  $\Phi_D$  are attributes. Triangles are labeled with their generating individuals.)

Consider for example the relation  $D$  of Figure 57. There are 17 individuals, each with three attributes. The figure also shows  $\Phi_D$ . We can see that there are no free faces, so the relation preserves attribute privacy by Lemma 61 on page 94. Moreover, each link  $\text{Lk}(\Psi_D, x)$  is homotopic to a circle  $\mathbb{S}^1$ . Indeed, viewed from attribute space, that link is exactly the boundary of a triangle for each individual. Figure 58 shows such a link for individual #10. The link relation has a large number of individuals but only three attributes. So Theorem 9 holds and there is homology in the link. There is however no homology in the attribute complex of the relation  $D$  itself; the simplicial complex  $\Phi_D$  is a triangulation of the Dunces Hat, a nontrivially contractible space.

Although  $R$  preserves attribute privacy, it does not preserve association privacy. Individuals 1 and 12 share exactly one attribute (namely  $h$ ), but do so with 4 additional people (namely 3, 7, 11, and 13). If attributes represent shared dinners, then in some cases one can infer all the guests at a dinner after having seen as few as two guests. (Attribute privacy means that one



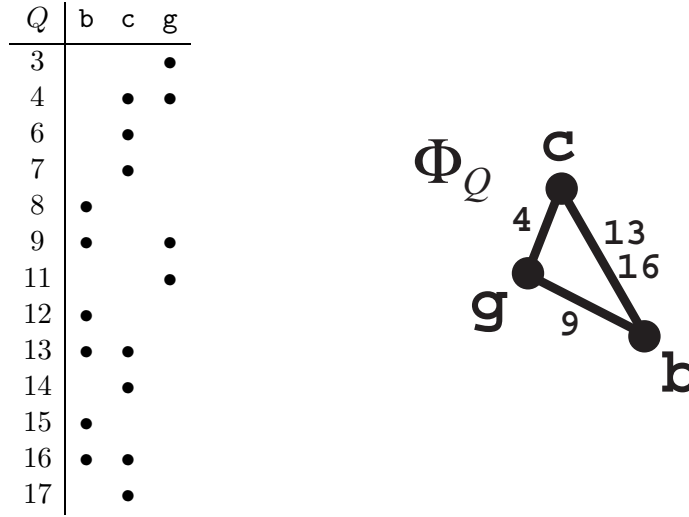


Figure 58: Relation  $Q$  represents  $\text{Lk}(\Psi_D, 10)$  from Figure 57. Also shown is the attribute Dowker complex  $\Phi_Q$ . It is the boundary of a triangle, so homotopic to  $\mathbb{S}^1 = \mathbb{S}^{k-2}$ . Since individual #10 has three attributes and  $1 = 3 - 2$ , that means relation  $R$  preserves attribute privacy for individual #10. (Vertices of  $\Phi_Q$  are attributes. Edges are labeled with their generating individuals. Notice that the edge  $\{b, c\}$  is generated by two individuals. Whereas most edges in  $\Phi_D$  are shared by only two triangles, edge  $\{b, c\}$  is shared by three triangles; it is one of those edges that are glued to two others in the Duncie Hat representation. — Individuals who generate just vertices are not shown in  $\Phi_Q$ .)

cannot infer additional dinners attended by a guest simply from having observed that guest at a particular dinner or two.)

## J.2 Disinformation

Privacy loss is possible when there is a free face in the relevant Dowker complex. One way to preserve privacy is to eliminate such free faces. Earlier in the report, we studied morphisms between relations as a possible way to transform data so as to reduce privacy loss. Ideally, for attribute privacy, the goal of such a transformation is to map onto a relation whose attribute complex has no free faces. We saw that such transformations need not always exist, for topological reasons, unless one is willing to introduce discontinuities, that is, discard knowledge of some relationships in the underlying spaces.

Alternatively, one could imagine embedding a relation within another that does preserve privacy. Of course, at the extreme, one simply embeds the given relation in a huge relation that looks like a perfect sphere. Now there is privacy but the same mechanism that provides privacy reduces utility. Nonetheless, one has not discarded relationships, merely surrounded them with disinformation. We saw an example of that early on, when we added a single attribute to relation  $H$  in the example of Section 3.1, in order to remove the inference that someone had cancer. If one has a separate mechanism for discerning fake entries from legitimate

entries, then one can see past the disinformation — in the earlier example that would entail having a (presumably safely encrypted) memory of which single entry in the relation is false.

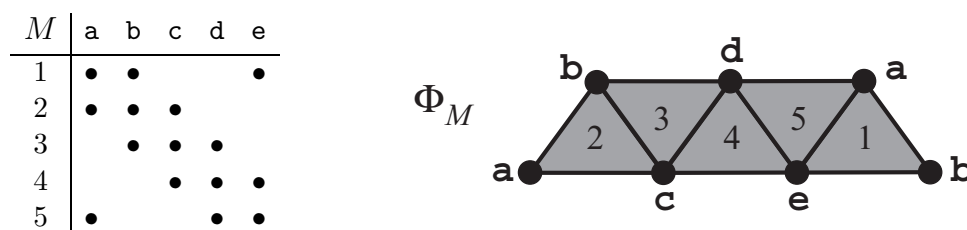


Figure 59: Relation  $M$  revisited along with its attribute complex  $\Phi_M$ .

Figure 59 revisits our earlier Möbius strip relation, showing the relation  $M$  and its attribute complex  $\Phi_M$ . Privacy loss occurs when someone observes two attributes that make up an edge on the boundary of the Möbius strip, such as  $\{b, d\}$ . Given the relation, the observer can infer a third attribute and identify the underlying individual, in this case infer attribute  $c$  and identify individual #3.

In order to preserve attribute privacy, one might consider adding decoy individuals whose so-called attributes include those edges, making them non-free, thus removing that inference mechanism. Figure 60 does so by doubling the number of individuals.

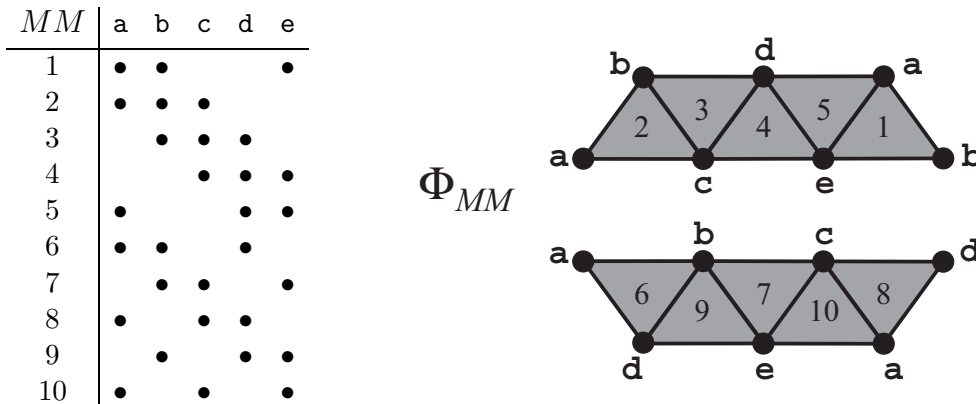


Figure 60: Relation  $MM$  adds five decoy individuals. The attribute complex  $\Phi_{MM}$  entails gluing two Möbius strips together.

The additional five individuals form their own Möbius strip. The figure therefore describes the overall attribute complex  $\Phi_{MM}$  as two Möbius strips, with edges shared between the two strips, as suggested by the vertex labels. The overall attribute complex amounts to gluing the two Möbius strips together, boundary to zigzag spine. The resulting attribute complex is the 2-skeleton of the full complex on the attribute set  $\{a, b, c, d, e\}$ . It therefore is homotopic to the wedge of four two-spheres:  $\mathbb{S}^2 \vee \mathbb{S}^2 \vee \mathbb{S}^2 \vee \mathbb{S}^2$ .

Each of  $\Phi_{MM}$ 's edges is now shared by three triangles. There are no free faces.

There is no attribute inference. (There is association inference.)

Moreover, the complex is sufficiently isotropic that one cannot say *a priori* which individuals are real and which are decoys, even if one knows that there might be decoys. Of course, the actual curator of the relation would need some secure mechanism to separate truth from fiction, that is, to peel apart the gluing. Regardless, real individuals may be identified upon seeing all their attributes (and only then).

### J.3 Insufficient Representation

In this subsection we show that if there are fewer than  $2^k$  individuals in a relation that has  $2k$  attributes representing  $k$  bits, then the relation cannot preserve attribute privacy for everyone. The basic reason is that fewer than  $2^k$  individuals amounts to cutting out a piece of the full attribute sphere  $\mathbb{S}^0 * \mathbb{S}^0 * \cdots * \mathbb{S}^0$ , exposing some free faces in  $\Phi_R$ . By similar intuition, it is *not* necessarily true that privacy loss will occur if there are fewer than, say,  $3^k$  individuals in a relation that has  $3k$  attributes representing  $k$  trivalent pieces of information. After all, bits are a special case of such tri-values, so one can preserve privacy with certain  $2^k$  individuals. Thinking topologically, the full attribute space for such tri-values looks like  $(\mathbb{S}^0 \vee \mathbb{S}^0) * (\mathbb{S}^0 \vee \mathbb{S}^0) * \cdots * (\mathbb{S}^0 \vee \mathbb{S}^0)$ . Cutting out a piece of that space does not necessarily create free faces, as one can see by simple example.

**Definition 94** (Binary Attribute Pair). *By a binary attribute pair we mean two mutually exclusive attributes, written  $y$  and  $\bar{y}$ . No individual can have both attributes. Moreover, in what follows we will assume that every individual has exactly one attribute from any such pair.*

**Lemma 95** (Privacy Requires Many Individuals). *Suppose  $Y = \{y_1, \bar{y}_1, y_2, \bar{y}_2, \dots, y_k, \bar{y}_k\}$ , with  $\{y_i, \bar{y}_i\}$  a binary attribute pair, for  $i = 1, \dots, k$ , and  $k \geq 1$ .*

*Let  $R$  be a relation on  $X \times Y$ , with  $X \neq \emptyset$  such that every individual  $x \in X$  has as attributes exactly one of  $\{y_i, \bar{y}_i\}$ , for each  $i = 1, \dots, k$ . Let  $n$  be the number of distinct rows of  $R$ .*

*Then  $R$  preserves attribute privacy if and only if  $n = 2^k$ .*

*Proof.* Observe that each row of  $R$  has exactly  $k$  nonblank entries, so each maximal simplex of  $\Phi_R$  consists of exactly  $k$  vertices. Moreover, no row of  $R$  is contained in another row unless the two rows are identical. We may therefore assume, without loss of generality, that all rows of  $R$  are distinct and incomparable. Consequently, every  $x \in X$  is uniquely identifiable. We can think of each individual  $x \in X$  as defining a unique  $k$ -bit number, with one bit per binary attribute pair, as determined by that individual's row,  $Y_x$ . All possible  $k$ -bit numbers are represented by  $X$  if and only if  $n = 2^k$ .

#### I. Suppose that $n = 2^k$ .

Showing that  $\Phi_R$  contains no free faces would establish that  $R$  preserves attribute privacy, by Lemma 61 on page 94. To show that  $\Phi_R$  contains no free faces, it is enough to show that, for every maximal  $\gamma \in \Phi_R$  and every  $y \in \gamma$ , the simplex  $\gamma \setminus \{y\}$  is contained in some maximal simplex of  $\Phi_R$  other than just  $\gamma$ .

Write  $\chi = \gamma \setminus \{y\}$ . Since  $y$  is a binary attribute pair, we can construct a new set  $\gamma'$  from  $\gamma$  by replacing  $y$  with its “opposite”. Specifically:  $\gamma' = \chi \cup \{y_i\}$ , if  $y = \bar{y}_i$ ; and  $\gamma' = \chi \cup \{\bar{y}_i\}$ , if  $y = y_i$ . Since  $n = 2^k$  there is an  $x \in X$  for which  $Y_x = \gamma'$ . So  $\gamma' \in \Phi_R$ , telling us  $\chi$  is not free.

II. Suppose that that  $R$  preserves attribute privacy.

By Lemma 62 on page 94,  $\Phi_R$  contains no free faces.

Let  $\gamma$  be a maximal simplex of  $\Phi_R$  and  $y \in \gamma$ . Define  $\chi = \gamma \setminus \{y\}$ . Construct  $\gamma'$  as in part I above. Consider the collection  $\Gamma = \{\eta \in \Phi_R \mid \chi \subsetneq \eta\}$ . The only possible set that might be in  $\Gamma$  besides  $\gamma$  is  $\gamma'$ . Since  $\Phi_R$  contains no free faces,  $\Gamma = \{\gamma, \gamma'\}$ .

Now vary  $y$  across  $\gamma$  and then repeat the process for all  $\gamma'$  thus constructed. The transitive closure of this process generates  $2^k$  distinct maximal simplices in  $\Phi_R$ , each of which corresponds to a unique  $x \in X$ . So  $n = 2^k$ .  $\square$

#### J.4 A Structural Inference Example : Passengers on Ferries

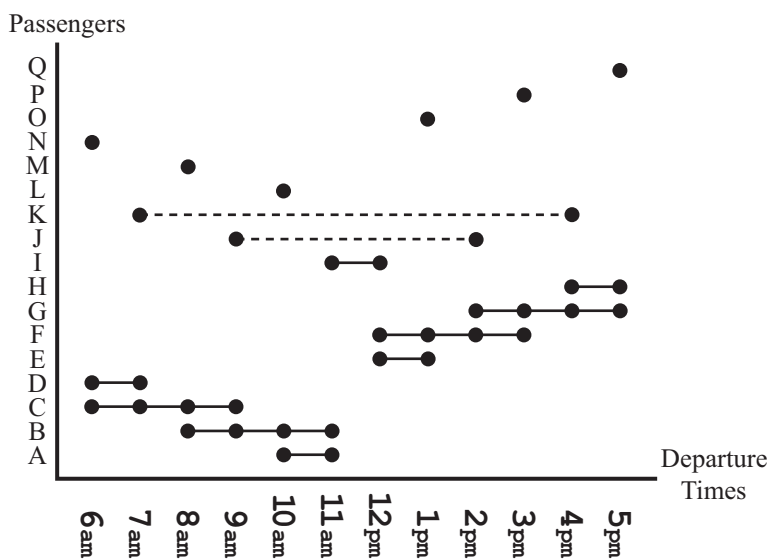


Figure 61: This time series represents 17 different passengers on 12 different ferry crossings. Each dot represents a passenger on a crossing. As a visual aid, solid lines connect multiple crossings by the same passenger at consecutive departure times, while dashed lines connect multiple crossings by the same passenger at non-consecutive departure times.

Imagine a commuter ferry that crosses back and forth between downtown and an island. Passengers pay electronically as they enter the ferry, so there is a record of who is on which crossing. Figure 61 shows a hypothetical time series for 12 crossings during a day in which 17 passengers took the ferry, some of whom crossed several times. Figure 62 shows the corresponding  $\Psi_R$  complex: vertices are individuals; each triangle represents a particular crossing. (Each ferry crossing had three passengers in this simplified example.)

The waitress in the ferry's coffee shop observes four individuals ordering coffee and conversing during the day, appearing in pairs on different crossings. She observes exactly four pairs, never the same pair twice. Who are the individuals?

It is convenient to represent the waitress's observations as a complex itself. Figure 63 does so. Vertices are now the four unknown individuals; edges are their (unknown) common crossing times. One can interpret who the individuals are by embedding the complex of Figure 63 into the complex of Figure 62, using one-to-one maps in both the passenger and time domains.

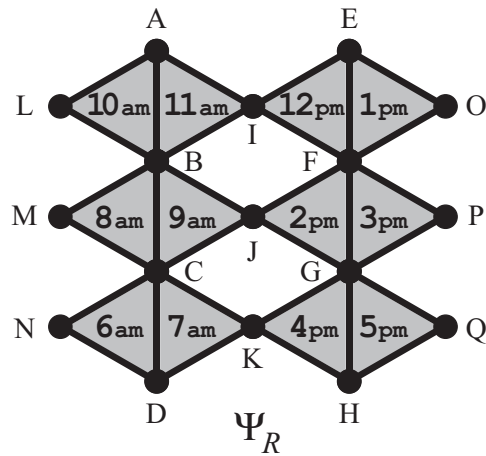


Figure 62: Simplicial complex  $\Psi_R$  determined by the time series of Figure 61, viewed as a relation  $R$ . Vertices represent passengers, labeled with letters. Triangles represent ferry crossings, labeled with departure times.

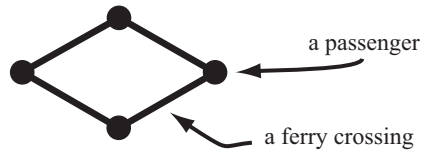


Figure 63: A waitress's observations of passengers drinking coffee together at various times, represented by a simplicial complex. Vertices represent unknown but distinct passengers. Edges represent unknown but distinct crossing times.

There are exactly two such embeddings (modulo index permutations), given by the two ways one can wrap a rectangle around the two holes in the complex of Figure 62. (Those are the only two “diamonds” touching four different crossing times.) Thus the individuals are either  $\{C, G, J, K\}$  or  $\{B, F, I, J\}$ , as indicated by Figure 64. Either way, one knows for sure that individual “J” twice had a conversation over coffee that day.

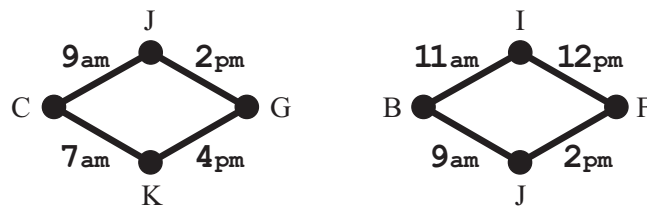


Figure 64: The two possible embeddings of the complex of Fig. 63 into the complex of Fig. 62.

Moreover, each of these embeddings places a time ordering on the embedded edges, from which one can make inferences as to who might have transmitted information to whom. For instance, for the embedding involving individuals  $\{B, F, I, J\}$ , one sees that individual “J” could have been both the initial source and final recipient of information.

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**Abstract**

Information has intrinsic geometric and topological structure, arising from relative relationships beyond absolute values or types. For instance, the fact that two people did or did not share a meal describes a relationship independent of the meal's ingredients. Such relationships give rise to lattices. Lattices have topology. That topology informs the ways in which information may be observed, hidden, inferred, and dissembled. Privacy preservation may be understood as finding isotropic topologies, in which relationships appear homogeneous. Moreover, the underlying lattice structure of those topologies has a temporal aspect, which reveals how isotropy may degrade over time, thereby puncturing privacy.

Dowker's Theorem establishes a homotopy equivalence between two simplicial complexes derived from a relation. From a privacy perspective, one complex describes individuals with common attributes, the other describes attributes shared by individuals. The homotopy equivalence is an alignment of certain common cores of those complexes, effectively interpreting sets of individuals as sets of attributes, and vice-versa. That common core has a lattice structure. An element in the lattice consists of two components, one being a set of individuals, the other being an equivalent set of attributes. The lattice operations join and meet each

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amount to set intersection in one component and set union followed by a potentially privacy-puncturing inference in the other component.

One objective of this research has been to understand the topology of the Dowker complexes, from a privacy perspective. First, privacy loss appears as simplicial collapse of free faces. The actual collapse is local, but the property of fully preserving both attribute and association privacy requires a global condition: a particular kind of spherical hole. Second, by looking at the link of an individual in its encompassing Dowker complex, one can characterize that individual's privacy via another sphere condition. That characterization generalizes to group privacy. Third, even when long-term privacy is impossible, homology provides lower bounds on how an individual may defer identification, when that individual has control over how to reveal attributes. Intuitively, the idea is to first reveal information that could otherwise be inferred. This last result in particular highlights privacy as a dynamic process. Privacy loss may be cast as gradient flow. Harmonic flow for privacy preservation may be fertile ground for future research.

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Topology Preserving Thinning of Cell Complexes.

IEEE Transactions on Image Processing,

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78:91-114, 2017.

doi:<http://dx.doi.org/10.1016/j.jsc.2016.03.009>

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